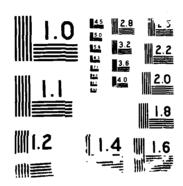
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A TIME SERIES ANALYSIS OF ENERGETIC ELECTRON FLUXES (1.2 - 16 MeV) AT GEOSYNCHRONOUS ALTITUDE

THESIS

Michael P. Halpin Major, USAF

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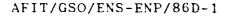
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#### THESIS

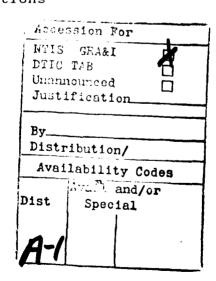
Presented to the Faculty of the School of Engineering
of the Air Force Institute of Technology
Air University
In Partial Fulfillment of the
Requirements for the Degree of
Master of Science in Space Operations



Michael P. Halpin, B.S.

Major, USAF

December 1986



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#### Preface

The person knowledgeable of the general subject of this thesis will perhaps notice a very marked similarity to two other theses accomplished at AFIT by prior students who also had an interest in the correlations between the solar wind, the interplanetary magnetic field (IMF), and geosynchronous energetic electron count rates (fluxes). These were the work of Captains Warren Smith and Douglas McCormick of previous GSO classes 83 and 84. In this study, new data were gathered and analyzed but, rather than using the classical methods of Smith and McCormick, the more refined method of Time Series Analysis as described by Box and Jenkins (1976) was used in an attempt to derive models for the energetic electron fluxes (Box and Jenkins, 1976). It is hoped that at least some of the models derived as a result of this thesis may be used to more successfully predict the level of energetic electron activity with sufficient lead time to allow efficient control of the many satellites we have operating in this energetic particle-filled region of space. Moreover, this thesis has been another attempt to show that some relationship exists between the IMF, the solar wind, and the energetic electron fluxes.

Perhaps the greatest thing one learns from taking on a project such as this is not just simply a few facts about the subject matter. On the contrary, what one "learns" for certain is the virtually ingraspable enormity of physical processes at work in the universe for which God has given us only a few keys to try to understand. In short, I have indeed had a humbling experience in composing this study.

Even more humbling is the realization of the incredible amount of stored knowledge that those who have helped me in this endeavor command at their mere fingertips. Major Joe Litko is a brilliant man whose near instantaneous and extremely accurate recollections of the Box and Jenkins time series details never ceased to amaze me. Concerning Major Jim Lange, I can only sing similar praises. His knowledge of the physics of the space environment is astounding. I am indeed greatly indebted for the continuing patience and help of these two individuals. I am also obliged to give my thanks to Mr. Jim Ware, the AFIT School of Engineering computer consultant for his much needed help in transfering raw data from tapes to files in my personal computer accounts. This enabled me to comfortably analyze the data in ways I am used to. Anyone who has

had the misfortune of having to deal with reading data off a tape knows what I'm talking about. Mr. Ware made the data reading process a virtual ease for me with his great command of the computer systems at AFIT. Thanks also goes to Dr. Ray Klebesadel of the Los Alamos National Laboratory and Ms. Billie Dolen and Mr. Ralph Post of the National Space Science Data Center for their very courteous and concerned help in providing the raw data for this project.

On the home front, where would I be without the comfort and restoration of my loving wife, Jennifer?

Even in my darkest moments she was able to help me keep my eyes focused on the goal and gave me reassurances that I was capable of completeing this project.

Finally, thanks be to the Lord. May I ever and always be His instrument and a part of His plan for good works.

Michael P. Halpin

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#### Abstract

This project used a Box and Jenkins time series analysis of energetic electron fluxes measured at geosynchronous orbit in an effort to derive prediction models for the flux in each of five energy channels. addition, the technique of transfer function modeling described by Box and Jenkins was used in an attempt to derive input-output relationships between the flux channels (viewed as the output) and the solar wind speed or interplanetary magnetic field (IMF) north-south component, Bz, (viewed as the input). The transfer function modeling was done in order to investigate the theoretical dynamic relationship which is believed to exist between the solar wind, the IMF B2, and the energetic electron flux in the magnetosphere. models derived from the transfer function techniques employed were also intended to be used in the prediction of flux values.

The results from this study indicate that the energetic electron flux changes in the various channels

are dependent on more than simply the solar wind speed or the IMF Bz. Also, most of the time series models developed here (for both the individual energetic electron channels by themselves and those developed through transfer functions) were not suitable for use in prediction, since the standard error of the forecasts made using these models was unacceptably high. However, a few of the models did merit possible consideration for use in prediction of fluxes. These were the individual time series models for the 6.6 - 9.7 MeV channel. In addition, the transfer function models developed using the solar wind as an input and the 6.6 - 9.7 MeV channel as an output may be of possible use. The channel containing electrons with energies between 9.7 - 16 MeV was also related to the solar wind via a transfer function with a reasonable forecast standard error. Finally, most of the transfer function models derived with the solar wind considered as the input to a given channel resulted in delay parameters of about 2 days between the input change in solar wind velocity and the observed output change in electron flux which supports findings from prior studies.

# A TIME SERIES ANALYSIS OF ENERGETIC ELECTRON FLUXES (1.2-16 MeV) AT GEOSYNCHRONOUS ORBIT

#### I. Introduction

#### Background

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In the years immediately after World War II and the capture of various German secret plans for rockets from their launch complex at Peenemunde on the Baltic coast, the imagination of the scientific community began to be taken with the possibilities for space exploration inherent in such devices as the V-2 which could zoom to altitudes many miles higher than had ever been previously achieved. Indeed, Dr. James Van Allen of the University of Iowa conducted a 1958 experiment with a V-2 spinoff, the Jupiter, which like its parent was developed by von Braun (Boyd, 1974:28). Van Allen put together a small package for the Explorer I satellite ir. which he installed Geiger counters in the Jupiter's payload and then helped launch it into space to continue the high altitude cosmic ray studies which he was then conducting (Glasstone, 1965:543). To Van Allen's surprise, he found that the term "space" apparently was a mishomer at least in the near-earth sense. The Geiger counters

recorded very high amounts of energy which caused a great bewilderment in the scientific community as to the energy's source. The <u>Van Allen belts</u>, as they are now called, are regions of very intense energy particles which extend from altitudes about 500 miles above the earth's equator to over 40,000 miles. Thus, man began to realize that this area of "space" (the magnetosphere) isn't just empty but is literally filled with these tiny, high energy particles.

But what are these particles, and where do they come from? A good portion of them are energetic electrons and protons which have become trapped in the earth's magnetosphere. It is believed that many of these particles are produced by solar disturbances, carried outward from the sun by the solar wind, and then deposited into the magnetosphere by a little understood mechanism which allows the Interplanetary Magnetic Field (IMF) to "link" up with the earth's magnetic field (Akasofu, 1983:179). Thus, the particles of the solar wind are allowed to enter the magnetosphere where they are subsequently accelerated and become "trapped" along the earth's magnetic field lines. The kinetic energy of these particles is very large which implies that they are extremely fast moving. The energies of the particles which will be dealt with in this study are in

the range of 1 to 16 MeV. One MeV is 106 electron volts where an electron volt is equal to approximately 1.6 x 10-19 joules of kinetic energy. Taking into consideration the very small masses involved, it should be apparent that the velocities of these protons and electrons are extremely high in order to produce energies of the magnitudes cited above. Indeed, with regard to the energetic electrons which will be studied here, the electrons must be considered in a relativistic sense, since their velocity may approach and in some cases nearly equal the speed of light (3 x 108 meters/sec).

The next question concerns what the implications are for our operations in space given all these particles in the near earth space environment. The implications are many. They can, for example, cause difficult problems for our spacecraft deployed in geosynchronous earth orbits inside the magnetosphere. High levels of flux (defined as the measured number of particles passing through a unit area per unit of time) of these small energetic masses can cause such maladies as: (1) spurious event sensing of surveillance satellites, (2) uncommanded tumbling motions of the satellite, (3) degraded sensor and/or electronics capabilities due to particle energy absorption over a

long period of time, (4) alterings of satellite surface coatings such that the satellite's operating temperature is increased above the optimum, and (5) radiation sickness, cancer, or gene mutation for astronauts traveling for any extended periods in such conditions (Lange, 1982:III-D-21,22; Spjeldvik and Rothwell, 1983:113).

Since it is imperative that we have satellites operating within the Van Allen belts in the geosynchronous region of space (approximately 22,700 miles altitude) for communications, surveillance, and exploration, and since the exact mechanism for the changes in flux of these energetic particles in the Van Allen belts is not well understood, statistical methods are used in an attempt to help explain particle behavior. Having some ability to predict when changes in flux levels might occur is a very desirable goal since such a capability would allow (given sufficient lead time) more efficient control of satellites which might be exposed to potentially harmful flux levels. For example, in the event of a forecast particle substorm, sensitive satellite subsystems might be turned off so as to avoid the deleterious effects of increased flux levels much as a personal computer owner might turn off his system at the onset of a thunderstorm to avoid

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the potential for damage due to lightning strikes.

Thus, many statistical studies of these energetic particle fluxes have focused on classical methods such as correlation analysis or analysis of variance (ANOVA) in attempts to link measured particle flux levels (recorded by a number of satellites which the US currently has in geosynchronous orbit) with other magnetospheric parameters which are also measurable by our satellites or at ground stations. The idea is that if a given magnetospheric or solar wind parameter can be used with sufficient lead time (say one to two days) to predict an increase in particle flux, then the parameter may be used as a basis for satellite command and control decisions.

One of the key areas of interest has been the belief on the part of many geophysicists that there is a relationship between changes in the solar wind velocity, the north-south component of the solar wind (or interplanetary) magnetic field (IMF), and the levels of flux of energetic electrons in the magnetosphere (Paulikas and Blake, 1978:2,10). Studies of note which have made attempts to correlate such parameters via the classical methods mentioned above are those of Smith (1983) and McCormick (1984). Smith, employing data from June 1979 to April 1982, used a correlation analysis and

ANOVA of solar wind speed, IMF, and energetic electron flux level data and concluded that there was only a weak correlation (largest value of R2 = 0.2582 when two-day old solar wind data used) between solar wind speed readings and energetic electron flux levels (Smith, 1983:42-50). Also, no significant correlations between the IMF north-south (B<sub>z</sub>) component and e-flux levels were noted (Smith, 1983:54). McCormick, with the same data set, used methods similar to those of Smith and came to virtually identical conclusions, but he also found that solar wind correlations with increased flux levels were slightly higher when the IMF north-south component was negative (ie, more southward oriented). To elaborate, R2 values were 0.269 and 0.313 respectively for the correlations between squared two day old solar wind speed values and energetic electron flux values from the two lowest electron flux channels of 1.2 - 1.8 MeV and 3.4 - 4.9 MeV. These  $R^2$  values were attained only when the IMF north-south (Bz) component was negative during the same time at which the solar wind speed was measured (McCormick, 1984:51-63). This tended to support existing ideas that a southward oriented Bz component of the IMF seems to result in more enhanced flux levels (Potemra, 1983:279).

Despite these sometimes discouraging results, the scientific community continues to believe that some sort of relationship exists between the IMF B<sub>2</sub>, the solar wind, and the energetic electron fluxes, and further study in this area seems warranted (Paulikas and Blake, 1978:1). In fact, a number of other studies have very much established that a statistical relationship does exist when electron fluxes of lower energy are involved (Su and Konradi, 1979:25).

#### Objectives of the Research

A comparatively new and unused method of statistical analysis for data taken over a long time period is that of <u>Time Series Analysis</u> such as that described by Box and Jenkins (1976). With the advent of very powerful computers and the availability of statistical programs such as the BMDP Statistical Software package, the capability to do analyses like that of Box and Jenkins now exists. Therefore, the objectives of this research will be twofold. First, a Box and Jenkins time series analysis of energetic electron flux data will be performed in an attempt to define models which may be used to predict changes in flux levels. As stated earlier, such models might allow the prediction of potentially dangerous fluxes with

sufficient lead time to allow better control of affected spacecraft. Second, since one of the main concerns of the geophysical/astrophysical community is in the linkage between the solar wind velocity, the north-south (Bz) component of the IMF, and the changes in flux levels of energetic e-, an attempt will also be made in this study to relate these entities via time series transfer functions. The scope of this study will be limited to data taken in the time period between April 1982 and May 1986, a period somewhat longer than that in the studies by Smith and McCormick. It should be emphasized that some of this data may be discarded due to the way in which a time series analysis is conducted. The finer details of a time series analysis as described by Box and Jenkins will be presented in the Detailed Methodology section of this report.

#### Overview

This report includes a literature review of articles on geosynchronous particle flux and its implications (Chapter II), data preparation and use of computers (Chapter III), a detailed look at the Box and Jenkins time series analysis (Chapter IV), a presentation of results (Chapter V), and finally, appropriate conclusions and recommendations (Chapter VI).

#### II. Literature Review

One of the most obvious points concerning a literature review in the general field of magnetospheric phenomena is the abundance of articles. This is due in part to the commissioning in the mid-1970s of the International Magnetospheric Survey (IMS) "in which a coordinated effort was made to understand magnetospheric processes" (Russell and Southwood, 1982:vii). was created by a group of concerned geophysicists who felt that more research was needed concerning the magnetosphere. The coordination of the many projects resulting from the multinational commitment to the IMS has been and continues to be carried out at the level of the participating scientists from each nation. result of the IMS has been a virtual (and much needed) flood of research projects and interest in the magnetosphere and its mechanics.

One of the best publications for finding literature on magnetospheric processes is the space physics portion (blue colored volumes) of the <u>Journal of Geophysical Research</u>. This is a monthly publication which is filled with the studies of the world's leading space physicists. Many of the articles cited here came from this journal.

The way in which one chooses to conduct a literature review of the magnetosphere is dependent on the specific project concerns. In this case, it was decided to look at articles and books associated with the general theme of the interactions between the IMF, the solar wind, and the flux levels of energetic particles observed in the magnetosphere. Also, were it not for the fact that these energetic particles pose difficult problems for our spacecraft, one might easily surmise that the intensity of research in this area would be somewhat lessened. Consequently, reviews of articles concerning the difficulties of space operations in the magnetosphere along with ways of predicting hazardous conditions there also seem appropriate.

#### Magnetospheric Interactions

A generally accepted theory is that the particle energy flux in the magnetosphere is greatly controlled by an as yet undetailed mechanism whereby the interplanetary magnetic field lines link with the earth's magnetic field lines. According to Nishida, extensive examinations of both ground based and space based observations have led us to believe that the energy supplied to the magnetosphere "proceeds mainly by reconnection between the lines of force of the IMF and

the geomagnetic field" (Nishida, 1983:185). When this occurs, vast amounts of energy in the form of particles carried along by the solar wind are allowed to enter the magnetosphere where subsequently these particles are accelerated to extreme speeds and become trapped along the earth's field lines. Paulikas and Blake, in an earlier study of 11 years of data on electron fluxes at geosynchronous orbit, found that the "efficiency of coupling" between these solar wind particles and the magnetosphere is apparently controlled by the "IMF direction as well as the solar wind velocity" (Paulikas and Blake, 1978:2).

Part of the problem in developing a deeper understanding of the processes controlling the rate of energy inflow to the magnetosphere has to do with the vastness of space itself. According to Baker, the very great distances involved make it extremely difficult to "probe the system concurrently at enough different points to truly understand the complex relationships between its different parts" (Baker, 1982b:5917). Su and Konradi, in an earlier paper, were in agreement and stated that "the observations made by a single spacecraft so far fail to resolve the temporal and spatial variations of the environment" (Su and Konradi, 1979:23). Nevertheless, by correlating data from

different satellites located in different parts of the magnetosphere and by systematically moving these probes so as to sample different areas, the work of understanding the processes continues. An example of such a study was that conducted by Baker in which he and co-workers developed a model of energetic particles at geosynchronous orbit by studying and analyzing the data sent back from six different satellites at different points in the magnetosphere during the occurence of a substorm on July 29, 1977 (Baker, 1982b).

Regardless of the actual mechanisms controlling the rate of energetic particle inflow, it is widely held that the solar wind is the driving force behind the inflow and holds the secrets to further understanding. In a recent study of high energy magnetospheric protons, Baker et al noted from their work that increases in energetic electron intensities (above 0.2 MeV) track closely with the solar wind velocity (Baker, 1979:7149). They also noted that virtually all substorms are accompanied by some observable injection of electrons with energies > 30 KeV. In a different article, Baker stated flatly that the dynamics of the magnetosphere "may be effectively discussed in terms of energy input from the solar wind into the magnetosphere" (Baker,

1982b:5917). Nishida again echoed this feeling in a later report (Nishida, 1983:185).

Still another facet in understanding the interactions is the part played by the north-south (B2) component of the IMF. Numerous studies have shown that increases in energetic particle flux seem to correlate well with a negative (or southward oriented) B2. Researchers involved in such studies included McPherron wno in observations of substorms found that the growth phase of these storms occurred during southward oriented  $B_z$  components of the measured IMF (McPherron et al, 1973:3131). Studies conducted by Russell in the following year came to the same conclusions (Russell et al, 1974:1108). Moreover, a 1977 study by Caan and others again reached the same conclusions regarding the onset of substorms following a prolonged (two hours or more) southward turning of the IMF B2 (Caan et al, 1977:4837). More recently, statements by Akasofu and Baker on this subject only serve to emphasize the unanimity of the geophysical community concerning the appropriateness of this theory. In a 1983 paper, Akasofu wrote,

. . . it has been found that the north-south component of the solar wind (or interplanetary) magnetic field is one of the most important parameters which link solar wind disturbances and magnetospheric disturbances. . . Thus, it is

crucial to understand the physical causes of theta (the north-south component of the solar wind magnetic field) and its time variations in linking solar wind disturbances and magnetospheric disturbances (Akasofu, 1983:179,181).

Baker was more to the point when in a later article that same year he stated,

. . . a southward IMF ( $B_z < 0$ ) gives rise to enhanced magnetospheric activity, while a prolonged northward IMF ( $B_z > 0$ ) is followed by quiet geomagnetic conditions (Baker and others, 1983:6230).

There are many more geophysicists in the available articles who are of the same beliefs.

Another point is also brought out in the literature and bears mentioning here. In a study to determine the effects of the solar wind on magnetospheric dynamics, Paulikas and Blake noted that increases in the solar wind velocity as well as the aforementioned southward oriented IMF correlated with increases in the numbers of energetic electrons in the magnetosphere (Paulikas and Blake, 1978:2). In fact, they were so adamant regarding the significance of the solar wind velocity that they went on to state that the "velocity of the solar wind is the most important parameter in organizing the flux levels of energetic electrons in the outer magnetosphere" (Paulikas and Blake, 1978:16). As part of their study they also

showed that increases in the flux levels of energetic electrons ( > 3.9 MeV ) correlated well with approximately 2 day old solar wind speed data. Specifically,

Starting about a day or two after the solar wind stream first reaches the earth, the fluxes build up in correlation with the increasing velocity of the solar wind (Paulikas and Blake, 1978:11).

This is in general agreement with the findings of Smith and McCormick who, as noted in Chapter 1, found some amount of correlation between two day old solar wind speed data and increased flux levels. (Smith, 1983 and McCormick, 1984). One interesting side light to the Paulikas and Blake paper has to do with their observation that the most marked variability of higher energy electron fluxes is associated with the 27 day solar rotation period (Paulikas and Blake, 1978:22). This is a characteristic which will clearly present itself when the results of the analysis are presented later on in this report.

Thus one can summarize the literature that surrounds the interaction aspect with three observations: (1) the solar wind is the energy provider, (2) the IMF linkage with the geomagnetic field is the key which apparently turns the energy flow on and off, and (3) the solar wind velocity, perhaps more than any

other parameter, is a good indicator of the amount of energy available for inflow to the magnetosphere.

#### Operational Hazards

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In a 1979 study, Grajek and Mcpherson stated that "a significant number of satellite operating anomalies are due to differential charging of spacecraft surfaces and resultant discharges" (Grajek and McPherson, 1979:769). They went on to observe that though most of these anomalies have had little impact on the spacecraft mission, some have been serious enough to result in total failure of the spacecraft power system (Grajek and McPherson, 1979:769). A separate study of the Van Allen belts in 1983 detailed some of the possible anomalies such as "detector malfunction and degradation, optical system degradation, memory system alteration, and control system malfunction or failure" (Spjeldvik and Rothwell, 1983:113). The study also cited the possible biological effects and implications for manned space operations (Spjeldvik and Rothwell, 1983:113). The main measure of damage done by penetrating energetic radiation or particles is radiation dosage or rads which is a unit of energy defined as the deposition of  $6.25 \times 10^7$  MeV in 1 gram of material. According to Spjeldvik and Rothwell, one of

the major concerns is "the on-orbit lifetime of microelectronic devices that are designed to a specific level of radiation 'hardness' (such as 104-105 rad)" which dictates a "trade-off between orbit choice and system lifetime" (Spjeldvik and Rothwell, 1983:114). Since just a single energetic electron of the type analyzed in this study may carry as much as 16 MeV, it is easy to see why space operations in the magnetosphere can not be taken lightly. In addition, the study noted that the upper limit for human tolerance to such radiation is only about 500 rads (lethal) while lesser amounts can cause serious biological damage (such as gene station or cancer) (Spjeldvik and Rothwell, 1983:121).

Therefore, given that man has found it necessary to operate spacecraft in the magnetosphere for a host of reasons, it will behoove us all to understand as much about magnetospheric processes as possible and to develop relationships for the prediction of such events so that sufficient lead time is available for spacecraft control. To a great extent, the practical approach up till now has been the observation of significant magnetospheric events and the correlation of these with measured geophysical parameters.

#### Prediction Methods

Numerous studies have been conducted for determining the correlations between events observed in the magnetosphere and measured parameters. An example of one of these studies attempted to prove or disprove previous correlations between such things as observed spacecraft anomalies and local time, geomagnetic activity, the day of the week, and the season (Grajek and McPherson, 1979). The study concluded that the anomalies are, with 99.997% confidence, dependent upon geomagnetic activity as measured by the index Dsr where DsT is a measure of the equatorial disturbance produced by magnetic storms (Grajek and McPherson, 1979:774). and Konradi also derived a model for particle flux intensities at geosynchronous altitude using a third order polynomial curve of best fit for their available data concerning particles of the energy range 70 -41,000 eV (Su and Konradi, 1979:27). During their study they also showed that geosynchronous particle flux intensities correlated well enough with the auroral electrojet (AE) index to indicate a definite causal relationship, while the flux intensities did not correlate well with the geomagnetic planetary (Kr) index (Su and Konradi, 1979:26,27).

Others have taken the approach of trying to directly include characteristics of the solar wind in their representations of magnetospheric processes. In a 1983 paper, Akasofu derived an equation for total energy output rate of the magnetosphere which contained, among other factors, the solar wind velocity and the square of the overall solar wind magnetic field magnitude (Akasofu, 1983:176). McCormick pointed out that this approach considered the magnetosphere to be a "driven" system rather than an "unloading system where substorms occur from energy accumulated in the magnetosphere and (are) released by some instability" (McCormick, 1984:16). Crooker and others after looking at six-month and yearly solar wind speed averages found a high correlation with geomagnetic activity and subsequently suggested that the product of the southward component of the IMF and the square (or higher power) of the solar wind velocity seemed to correlate well with geomagnetic activity (Crooker et al, 1977:1933-1936).

Alas, there are many theories concerning ways to predict particle flux in the magnetosphere, but as yet there is no universally acceptable way to predict, much less describe, magnetospheric processes. Perhaps Paulikas and Blake summed up the hopes of many space researchers when they stated:

. . . it seems clear that the present results already offer some hope of both short-term and long-term prediction of the energetic electron radiation in the outer magnetosphere if a sufficiently accurate prediction of the parameters of the solar wind are available (Paulikas and Blake, 1978:33).

With the hope of deriving prediction models for energetic electron flux in the magnetosphere based on the Box and Jenkins time series analysis methods along with the same hope for deriving a relationship between the flux and the IMF/solar wind speed via transfer functions, let us go on to consider how the data were prepared.

#### III. Data, Software, and Computers

#### Data

The data for this project were supplied by two sources. The Los Alamos National Laboratory and the satellite 1979-053 were the source of one set of the data. The National Space Science Data Center through readings from the IMP 8 and ISEE 3 satellites was the other.

The Los Alamos data was originally sent on a magnetic tape and contained, among ten variables, daily average values of five energetic electron flux level channels covering the time period from April 1982 to May The details of the orientation of satellite 1979-053 and its sensors which record the flux values are available from Baker et al. The five channels mentioned contain measurements for flux levels (count rates) in the ranges of 1.2 - 1.8 MeV, 3.4 - 4.9 MeV, 4.9 - 6.6 MeV, 6.6 - 9.7 MeV, and 9.7 - 16 MeV. Two different detector packages aboard the satellite collected this data. The lowest energy channel (1.2 -1.8 MeV) was collected by a solid state detector. Henceforth, this channel's data shall be referred to as the SEESSD data channel (where "SEE" is an acronym for spectrometer for extended electron measurements and

"SSD" stands for solid state detector) (Baker et al, 1982a:83). The last four channels were collected by the other onboard detector and will be referred to henceforth as the SEEI, SEEII, SEEIII, and SEEIV data channels respectively. One point which should be made is that the flux values measured are simply the count rates which are not the same as the standard units for flux levels (particles per unit area per unit time). However, the correlation between count rates and the standard units is direct, so the measured count rates may be used in an analysis just as the data converted to standard units could be used.

Regarding the second set of data supplied by the National Space Science Data Center (NSSDC), once again, the data were sent on a magnetic tape referred to commonly by the NSSDC as the "Omni Tape". This tape, in addition to containing the pertinent values for solar wind speed and the IMF components, contained readings of no less than 37 different parameters. The big difference between this data and that supplied by Los Alamos, however, was that the Omni Tape contained hourly readings of all these different parameters over a period stretching from April 1981 to April 1985. This meant that the data in its raw form was contained in a five megabyte file with much of the information irrelevant to

this study. Consequently, a computer program had to be written to allow compression of the original Omni Tape supplied into a more manageable file containing daily averages. This was done in order to synchronize daily average values for the solar wind speed and IMF with the daily average readings contained on the Los Alamos tape for the energetic electron fluxes. Though the NSSDC obtains their Omni Tape data from a total of 17 different satellites, the majority of the readings for the solar wind speed and the IMF component values come from the IMP 8 (for International Magnetospheric Probe) and ISEE 3 (for International Sun-Earth Explorer) spacecraft. The ISEE project is a joint NASA-European Space Agency effort to study the outer magnetosphere (von Rosenvinge, 1982:1). Once again, details concerning spacecraft orbital parameters and instrumentation are available from McCormick (McCormick, 1984:7) and King (King, 1982:10-20).

## Overview

As a first action, both tapes were read into personal files on the CDC Cyber 6000 and SSC Unix VAX 11/780 computers at AFIT. Next, due to the intractable nature of the data contained on the Omni Tape, its data was compressed into daily averages and then saved in a

different file. The Los Alamos data for the energetic electron fluxes (count rates) was already in this daily form. A separate file containing time synchronized solar wind speed, IMF B<sub>2</sub>, and energetic electron flux values was also created to allow transfer function modeling. Once this was accomplished, analyses of the data began. The BMDP Statistical Software package used in this study was available on both systems. Also, files were transferable between the two computers. The availability of both systems for this project helped reduce time delays when one system was busy. This was helpful in that many analyses were necessary with a great deal of iterative interactions on the part of the author.

# <u>Detailed Preparations</u>

As explained earlier, the data for this study were obtained from LANL and NSSDC in the form of tapes. The LANL tape contained the data for daily average count rates in five different energetic electron energy ranges (or channels). This tape was obtained first and was delivered to the AFIT school of engineering computer consultant who was able to successfully load the tape on the school's tape drive unit in the computer terminal room. Using a program to transfer data from a tape file

to a disk file in the SSC's memory, the consultant then transfered the data to a personal file on the SSC which was subsequently copied by the author to enable later analysis via BMDP. The data supplied by LANL are explained in more detail in an article by Baker (Baker, 1982a:82-90).

The data supplied by the NSSDC (Omni Tape) also came in magnetic tape form. The problem with this file was its great size (approximately 5.2 megabytes). process followed for "getting" the data from this tape was somewhat different from that followed for the LANL tape. The tape was loaded on the school's tape drive unit in the computer room. However, since its size was too great to be accommodated in a personal file, the consultant loaded it into a general purpose directory of the SSC's memory. This directory is purged every 48 hours, but its availability allowed the author enough time to write a small FORTRAN program which could access the data from the Omni Tape (as written in the general purpose directory) and then compress it into daily (24 hour) averages. Once this was done, the size of the new file was reduced by 1/24, and this made it manageable and small enough to be stored in the author's personal SSC account. All daily average parameters (from the original 37) not seen to be of any use to this study

were discarded in order to further reduce the size of the file. For completeness <u>program readom</u>, a listing of the FORTRAN program written to compress the Omni Tape, is included in Appendix A of this report. Adequate comments are available within the program listing to enable the reader to understand what was done to actually read and compress the NSSDC data. Details of the data contained on the Omni Tape are contained in a separate set of articles (von Rosenvinge, 1982:1-9 and King, 1982:11-20).

A set of particulars concerning each tape's generation also accompanied them in the mail. This was both helpful and necessary in each case, since without knowledge of the format of the information on each tape, the writing of a program to read the data would have been impossible.

In order to perform a time series analysis using transfer functions, it was necessary to write another FORTRAN program which could read values from the NSSDC file (solar wind/IMF values) and the LANL file (electron flux values) and then store these values in a single time synchronized data file. This was required so that computation of the cross correlation functions required in transfer function modeling could take place. Only the values from overlapping days of the two files were

read and stored, since the cross correlation between two time series only makes sense if the series are synchronized in time. Thus, progam combol read and stored 706 combined cases of data from May 8, 1983 to April 12, 1985. This encompassed approximately the last two years of the NSSDC data on solar wind speed and IMF and the middle two years of the LANL data on the energetic electron flux. This was more than sufficient to perform a suitable transfer function analysis. A copy of program combol is also included in Appendix B.

## Missing Values

Some important points concerning the data should be mentioned which have a direct bearing on this study. On both tapes some data were naturally missing. This is of course due to the inherent problems with trying to obtain satellite readings via sensors which can malfunction. Missing data can have serious implications for a time series analysis which may only be accomplished on a set of equally spaced (in time) readings. To remedy this problem, two different methods were used. Regarding the energetic electron data (the tape from LANL), missing values were estimated by BMDP program AM, "Description and Estimation of Missing Data" using the SINGLE method whereby "a value for a variable

is estimated by regressing that variable on the variable with which it is most highly correlated" (Dixon et al, 1985:217-234). This seemed like a logical method since the data contained five different electron flux channels with at least some amount of presumable correlation between the channels. The use of program AM was possible because there was a total of only 32 days over the four year period covered by the data where missing values occurred. Also, the smaller size of the LANL file made it easy to apply program AM. The majority of these missing values were in August of 1982 (17 were missing) which was subsequent cause for disregarding the data from 1982 for this study. The estimated values provided by program AM were then substituted into the data set to allow analysis.

The NSSDC data, however, presented a slightly different problem. The massive size of the Omni Tape made its data intractable as far as using BMDP program AM. Therefore, the author decided that the best way to handle missing data on the Omni Tape was to allow program readom to store the last computed daily average as the new daily average if an entire day's set of hourly readings was missing. Also, if only part of one day's hourly readings was available for averaging (say only 10 hourly readings instead of the usual 24 for one

day), then those readings which were available were averaged and taken to be the daily average value. Again, only in the case where an entire day's set of hourly readings was missing was the previous day's average substituted. With the data set filled in, the analysis could begin. The data files along with the BMDP instruction files were easily transfered back and forth between the SSC and the Cyber.

### Discarded Data

One final point should be made. As mentioned earlier in this chapter, only about the last three years of the data from the LANL tape were used. This was due to the fact that the LANL tape contained the majority of its missing values (17 out of 32) in the first year of the data (eg., in 1982). Thus, to reduce the effect of substitution of estimated missing values, this data was discarded. In addition, since the LANL data ended on May 6, 1986, it was decided to discard all data up to May 8, 1983 which allowed the analysis of precisely the last three years of the data (1095 days). It was felt that the analysis of the most recent data would allow the development of the most accurate and up to date time series models.

In the case of the Omni Tape, only the last two years were used for the transfer function modeling (May 1983 to April 1985) because this gave the maximum overlap of the LANL data. The reader will recall from earlier discussions that time synchronized data from two series is necessary in order to do transfer function modeling. The overlap occurred between the dates mentioned above and was more than adequate to perform the analysis.

With regard to the actual data which was used,

Appendix C is a FORTRAN program which was composed to

write out all the pertinent data utilized in this study.

Appendix D is the result of this program: a complete

listing of the data.

# Software and Computers

BMDP2T, the "Box-Jenkins Time Series Analysis", was the program used to perform all the analyses (Dixon and others, 1985:639-660). The initial step in running BMDP2T is to compose a small computer program containing the necessary instructions to perform a basic examination of the data (Dixon and others, 1985:640). The stored data files are accessed by either including them directly as part of the instruction file (as called for on the Cyber) or by accessing them external to the

instruction file via a data file identification statement (as called for on the SSC). The basic examination exhibits the data for each time series in a graph of the daily flux values over time (a time plot). In addition, it performs an initial calculation of the autocorrelation and partial autocorrelation functions of the data so as to help define an initial guess at a prediction model. Such terms as "autocorrelation function" along with the details of the Box and Jenkins time series analysis methods will be given later on in the Detailed Methodology chapter of this paper. For now, it is sufficient for the reader to understand that this type of analysis combines a certain amount of precision with an equal amount of "art" and "gut feeling". Said more formally, although this method does have some "hard rules", its application to real data involves some interpretation. Time series analysis is an iterative and time consuming method. The prediction models derived from a time series analysis are not obtained by a definite set of procedures. Moreover, there may be more than one model which is appropriate for the data depending on the total history or extent of the data which is used. Time series models are always subject to updating when new data becomes available to add to the history of the series.

Two computer systems were used in this research. The CDC Cyber 6000 provides the fastest processing capability regardless of the task (usually no more than 30 seconds), but it is a bit more restrictive than the SSC Unix system which has better manipulative and naming capabilities for files. The BMDP Statistical Software package is readily available on both systems. current BMDP manual is the 1985 reprinting (Dixon and others, 1985). Documentation for the BMDP programs as implemented on either system is available. For the SSC, a file named <a href="mailto:bmdp.doc">bmdp.doc</a> (documentation file) is made available to the user when the entire set of BMDP programs is initially accessed. This file contains helpful information concerning how to actually run a given BMDP program. Included are tips to increase memory size (RAM) alotted for each program so that all appropriate processing may be performed as necessary. This one particular feature was utilized often due to the size of the data sets involved along with the amount of processing. A file similar to the bmdp.doc file is also available for the Cyber implementation of BMDP. The on-duty computer consultant can obtain a hard copy of this file for anyone desiring it as well as show the user how to access the BMDP software on either the Cyber or the SSC.

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# IV. Detailed Methodology

# The Box and Jenkins Method

The book entitled Time Series Analysis: Forecasting and Control by G.E.P. Box and G.M. Jenkins (1976) is the backbone for the analysis in this report. The basic ideas in a time series analysis are presented in the first chapter of this book (Box and Jenkins, 1976:1-19). As compared to its more classical statistical counterparts, this method is relatively Therefore, a few of the fundamental ideas in unknown. time series analysis will be discussed along with some of the models which form a framework for a time series forecast function. In addition, the reader will be made aware of how a time series transfer function is derived. Those with a deeper interest in the workings of the Box and Jenkins method are strongly urged to obtain a copy of the text for their own perusal.

# Time Series

A series of recorded values of a given random variable taken at equally spaced intervals of time is a discrete time series. Since many entities in the world can be termed random variables, analysts thus have the opportunity to record periodic values of these entities

and study them as they behave over time (ie., in a time series).

A time series analysis has three important "areas of application" as mentioned in the introductory chapter. Two of these have direct bearing upon this study and are noted as

- (1) The forecasting of future values of time series from current and past values.
- (2) The determination of the transfer function of a system—the determination of a dynamic input—output model that can show the effect on the output of a system subject to inertia, of any given series of inputs (Box and Jenkins, 1976:1).

Recalling the introduction, the reader will note that the objectives of this study were to derive prediction (forecasting) models for energetic electron flux and also to derive relationships (transfer functions) between the IMF B<sub>2</sub>, the solar wind speed, and the energetic electron flux. This second objective is undertaken with the IMF or the solar wind speed viewed as the input series and the energetic electron flux in any given channel viewed as the output. As noted by Box and Jenkins, a good forecast function can provide the basis for correct planning and control (Box and Jenkins, 1976:1). In this particular case, a good forecast function would allow optimal control of spacecraft and sensors which may be exposed to predicted dangerous

levels of energetic electron flux. Also, a valid transfer function would help to establish more firmly the theoretical relationship between the IMF, the solar wind, and the energetic electron flux.

# Fundamentals

A random variable (rv) of interest whose values have been recorded over equispaced intervals of time might be designated as z. If we denote the present time as t, then z: represents the currently recorded value for the rv of interest. Similarly, one might denote the last recorded value of z prior to the current value as z:-1. The value recorded two time periods prior to the current would be  $z_{t-2}$  and so on. The elapsed time period between which the readings of the variable are made can take on any value. Often times, it will be dictated by the apparatus used to measure the variable (in this case, the time between satellite sensor measurements of the energetic electron flux, the IMF component values, and the solar wind speed). important point is that the individual discrete readings of the variable must be equispaced over time.

What is sought, then, is a forecast of the value of the variable at some future time t+1. One might denote this forecast as  $\frac{\Delta}{2}$ , (1) where  $\frac{\Delta}{2}$  (pronounced

zee-hat) is the estimate of the series' value at time t+1. According to Box and Jenkins, the objective is to derive a forecast function such that "the mean square of the deviations  $z_{t+1} - \hat{Z}_t(1)$  between the actual and forecasted values is as small as possible for each lead time 1" (Box and Jenkins, 1976:2).

In addition, it is necessary to specify the accuracy of each forecast function so that "the risks associated with decisions based upon forecasts may be calculated" (Box and Jenkins, 1976:2). These accuracies are typically specified by calculating probability limits (confidence intervals) on either side of each forecast value (Box and Jenkins, 1976:2). Thus, if one desired an estimate for the energetic electron flux two days in the future based upon the present history of the data, the forecast model employed might yield a value (with arbitrary units) of, say, 3.6 + - 1.2 where the 1.2 on either side of the estimated 3.6 might represent the standard error for the estimate. This being the case, the given estimate along with the stated +/-1.2standard error interval would represent a confidence interval (or probability limits) for the estimate of approximately 68%. A 95% confidence interval estimate would then be represented by (approximately) 3.6 +/-2.4. Obviously, if one desires a greater accuracy, then

the confidence interval for the estimate has to become broader. In addition, the farther into the future that a prediction is desired, the broader the confidence interval for such an estimate will be.

To draw a parallel between what has been discussed thus far and the data presented here for analysis, Figure 1 is shown below. In it is a plot

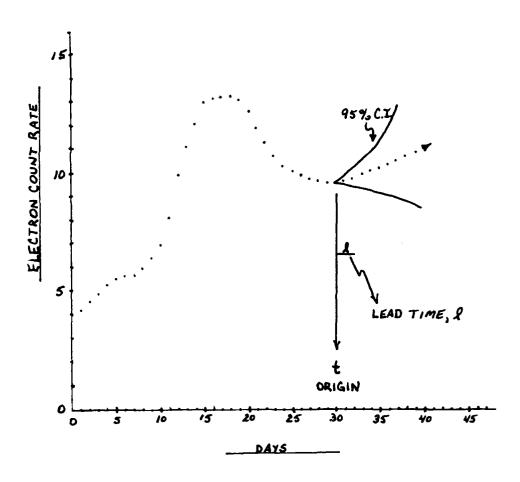


Fig 1. Hypothetical Time Series Plot (Box and Jenkins, 1976:2)

of a hypothetical time series for energetic electron flux readings recorded over a period of 30 days. If the present time is t, then what is desired is an estimate of the series' value at time t+1 where 1 is perhaps two days hence. Note that the confidence intervals shown illustrate that the farther into the future one estimates, the broader the confidence interval becomes. Likewise, the greater the specified accuracy of the estimate at any desired time in the future, the broader the interval.

Notice that the word "probability" is used in the above discussion. This is because a time series model is a probabilistic or stochastic model as opposed to a deterministic model. In a stochastic model, one calculates "the probability of a future value lying between two specified limits" (Box and Jenkins, 1976:7). It is also important to note that the actual time series recorded is just one of an infinite number of possible series which could have been generated by the given stochastic model for the series (Box and Jenkins, 1976:7).

#### Categories of Processes

Generally, there are two different classes of time series models. Stationary models derive their name

from the fact that the series they represent tend to be in equilibrium about a constant, never-changing, mean level (Box and Jenkins, 1976:7). A nonstationary model, however, is used to represent a series (also called a process) where there is no natural mean or equilibrium point (Box and Jenkins, 1976:7). As will be explained later, a given time series usually reveals whether or not it is stationary when its autocorrelation function is plotted. If the autocorrelations decrease quickly as the lag, k, increases, the series is considered stationary. In all the cases studied here, the series were stationary. A particularly wide class of processes which "provides a range of models, stationary and non-stationary, that adequately represent many of the time series met in practice" is known as the autoregressive-integrated moving average processes or ARIMA processes (Box and Jenkins, 1976:8).

# Operators and Assumptions

In describing the primary time series models, it is necessary to discuss a couple of the basic operators used. Perhaps the two most important are the backward shift operator denoted by B and the backward difference operator denoted by V (pronounced "del"). These two operators are defined as follows:

 $Bz_t = z_{t-1}$ 

and

 $B^{m}Z_{i} = Z_{i-m}$   $\nabla Z_{i} = Z_{i-1}$   $= (1 - B)Z_{i}$ 

and

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 $\nabla^m z_t = z_t - z_{t-m}$ 

Note that the backward difference operator can be represented in terms of the backward shift operator.

Also, the parameter m may be thought of as the "power" of the operator.

In addition to these basic operators, the time series models used by Box and Jenkins view the output of any given time series as being generated from an input series of independent shocks,  $a_1$ , with the shocks being random drawings from a fixed normal distribution of mean zero and variance  $\sigma_2^2$ . "Such a sequence of random variables  $a_1$ ,  $a_{1-1}$ ,  $a_{1-2}$ ,... is called a white noise process by engineers" (Box and Jenkins, 1976:8). When the model for the series under study is finally fitted, the white noise represents the <u>residuals</u> or the differences between the actual series values and those

which the model would predict at the same points in time at which the actual series values were recorded. This sequence of random white noise numbers is transformed into the observed output time series,  $z_t$ , by another operator known as a linear filter,  $\psi(B)$  where B is again the backshift operator and the notation  $\psi(B)$  indicates that varying powers of backshifting are used throughout the operator. For example, one might represent a time series of interest by the model

$$z_{t} = \mu + \psi(B)a_{t} \tag{1}$$

where the filter,  $\psi$ (B), is

$$\psi(B) = 1 + \psi_1 B + \psi_2 B^2 + \dots$$

A finite or infinite/convergent set of parameters  $\psi_1$ ,  $\psi_2$ , ... is said to classify the process as stationary. In this case, the value  $\mu$  is thus the natural mean or equilibrium point for the process under study. If the set of parameters is infinite and not convergent however, then the process is said to be nonstationary, and in this case,  $\mu$  is only a "reference point for the level of the process" (Box and Jenkins, 1976:8-9).

# Specific Models

An <u>autoregressive</u> model for a time series is expressed "as a finite, linear aggregate of previous values of the process and a shock a<sub>t</sub>" (Box and Jenkins, 1976:9). If the values of a process are  $z_t$ ,  $z_{t-1}$ ,  $z_{t-2}$ , .... as before and the deviations of each of these values from the mean,  $\mu$ , are represented by  $\tilde{z}_t$ ,  $\tilde{z}_{t-1}$ ,  $\tilde{z}_{t-2}$ , .... then an autoregressive (AR) model of order p is represented by

$$\tilde{z}_{t} = \phi_{1}\tilde{z}_{t-1} + \phi_{2}\tilde{z}_{t-2} + \ldots + \phi_{p}\tilde{z}_{t-p} + a_{t}$$
 (2)

which as the reader will note, is simply a regression equation where the value  $\mathcal{Z}_{i}$  is regressed on previous values of itself. The term  $a_{i}$  is analogous to the error term in a regression and in this context represents the random shock input in the current time period. Thus, the reason for the name <u>autoregressive</u> becomes apparent (Box and Jenkins, 1976:9). The autoregressive operator may then be defined as

$$\phi(B) = 1 - \phi_1 B - \phi_2 B^2 - \dots - \phi_p B^p$$

and in doing so, the autoregressive model above may be rearranged and "written economically" so that

$$\phi(B)\tilde{z}_t = a_t$$
 (Box and Jenkins, 1976:9)

As Box and Jenkins point out, it is not difficult to see that the autoregressive model, (2), is a special case of the linear filter model, (1), discussed previously (Box and Jenkins, 1976:10). To show this, the z terms on the right hand side of (2) may be progressively replaced with their expanded values according to the model. For instance, the term  $\mathbf{Z}_{t-1}$  on the right side of (2) may be replaced and thereby eliminated by substituting

$$\tilde{z}_{t-1} = \phi_1 \tilde{z}_{t-2} + \phi_2 \tilde{z}_{t-3} + \dots + \phi_p \tilde{z}_{t-p-1} + a_{t-1}$$

Similar substitutions for  $\tilde{\mathbf{z}}_{t-2}$ , and so on will eventually yield an infinite series in the a's (Box and Jenkins, 1976:10). Having made all the substitutions, one can thus see that not only is

$$\phi(B)\tilde{z}_{t} = a_{t}$$

as already pointed out, but also

$$\mathbf{z}_{t} = \boldsymbol{\psi}(\mathbf{B})\mathbf{a}_{t} \tag{3}$$

Therefore, it is obvious that

$$\psi(B) = \phi^{-1}(B)$$
 (Box and Jenkins, 1976:10)

The reader will note that the utility of using  $m{\phi}(\mathtt{B})$  over  $m{\psi}(\mathtt{B})$  in model development stems from the fact that while

 $\psi$ (B) has an infinite number of terms,  $\phi$ (B) is economically represented by just p terms.

Whereas the autoregressive model could express  $\mathbf{Z}_{t}$  as the sum of an infinite series of weighted shock values shown in Eq (3) above, the moving average model expresses  $\mathbf{z}_{t}$  as the sum of a <u>finite</u> series of weighted shocks. The model is thus expressed as

$$\tilde{z}_{t} = a_{t} - \theta_{1}a_{t-1} - \theta_{2}a_{t-2} - \dots - \theta_{q}a_{t-q}$$
 (4)

and is called a <u>moving average process of order q</u> or MA.

The MA operator is thus defined as

$$\theta(B) = 1 - \theta_1 B - \theta_2 B^2 - \dots - \theta_q B^q$$

and the MA model may thus be written more succinctly as

$$\tilde{z}_t = \theta(B)a_t$$
 (5) (Box and Jenkins, 1976:10)

The mixed autoregressive-moving average (ARMA) model is used to achieve greater flexibility in the fitting of models to a given time series (Box and Jenkins, 1976:11). This model consists of the addition of the terms from the right side of Eq (4) to those from the right side of Eq (2) to yield

$$\tilde{Z}_{t} = \phi_{1}z_{t-1} + \ldots + \phi_{p}z_{t-p} + a_{t}$$

$$- \theta_{1}a_{t-1} - \ldots - \theta_{q}a_{t-q}$$
 (6)

or

$$\phi(B)z_t = \theta(B)a_t \tag{7}$$

For discrete data such as this study will analyze, the <u>transfer function</u> between an output time series, Y<sub>1</sub>, and an input time series, X<sub>1</sub>, may be effectively represented by a difference equation involving backward shift operators which operate on both the input and the output series (Box and Jenkins, 1976:13). This type of representation is given as

$$(1 - \delta_{1}B - \dots - \delta_{r}B^{r})Y_{t}$$

$$= (\omega_{0} - \omega_{1}B - \dots - \omega_{s}B^{s})X_{t-b} + a_{t}$$

$$= (\omega_{0}B^{b} - \omega_{1}B^{b+1} - \dots - \omega_{s}B^{b+s})X_{t} + a_{t}$$

$$(8)$$

or

$$\delta(B)Y_t = \omega(B)BbX_t$$

$$= \Omega(B)X_t$$

where the parameter b is the number of time periods (dead-time or pure-delay) between the input value and its observed effective output. The parameters r and s express the order of the left side and right side operators in the above difference equation and at is a random noise component. Another way of stating the

relationship is that  $Y_t$  and  $X_t$  are linked by a linear filter such that

$$Y_t = v_0 X_t + v_1 X_{t-1} + v_2 X_{t-2} + \dots + a_t$$
 (9)

or

$$Y_t = v(B)X_t + a_t$$

where v(B) is the <u>transfer function</u> and can be stated more explicitly as

$$v(B) = v_0 + v_1B + v_2B^2 + \dots$$
 (10)

Thus v(B) is simply a ratio of the right side operator of order s and the left side operator of order r in the difference equation (8). The reader should take note of the fact that for an input-output relationship in which the dead-time between the input and its observed output is equal to b time periods, the first b weights in equation (10) (eg.,  $v_0$ ,  $v_1$ , ...,  $v_{b-1}$ ) are zero (Box and Jenkins, 1976:14).

With regard to the data available for this study, it is hoped that a model of the types discussed here will fit. The tasks thus implied with respect to the objectives listed in the introduction to this report are as follows:

- (1) Develop, if possible, a time series model of the type AR, MA, or ARMA for each of the energetic electron channels discussed.
- (2) Use the developed models to make forecasts of the flux values and compare these forecast values to the actual values to assess the model's utility in prediction.
- (3) Develop models of the same type for solar wind and IMF  $B_z$  in an attempt to show that the solar wind speed or the IMF  $B_z$  (considered as inputs) are related to the energetic electron flux levels (considered as outputs) via a transfer function.

## Deducing the Model

The "precise steps" to be followed in deriving these models have so far not been mentioned. As stated in the Chapter III, the Box and Jenkins method combines a certain amount of precision with an equal amount of art. The "precision" in deriving the appropriate model(s) comes with the determination of the autocorrelation and partial autocorrelation functions. The "art" comes in the form of how well one can deduce the type of model from these special functions.

Deducing the model is an iterative and time consuming process which involves close scrutiny of these functions as many models are fitted to the data until eventually (and hopefully) one of the fitted models leads to a reasonably small set of residuals.

Thus, a necessary first step is the determination of the autocorrelation functions for the

time series of interest. For a series with N discrete values, the kth lag autocorrelation is defined as

$$r_k = C_k/C_0 \tag{11}$$

where

$$C_{k} = -\sum_{\substack{N = 1 \\ N = 1}}^{1 - N - k} (z_{t} - \overline{z})(z_{t+k} - \overline{z}), \quad k = 0, 1, 2, ..., K \quad (12)$$

is an estimate of the autocovariance at lag k, and  $\overline{z}$  is the mean of the time series (Box and Jenkins, 1976:32). The lag (k) has to do with the number of time periods between recorded values in the process. For instance, if the series of interest consists of 1000 daily readings (eg., readings taken precisely 24 hours apart), then the autocorrelation at lag k = 2 days would consist of the sum of all autocovariances of readings which lag each other by exactly two days (C2) divided by the sum of all autocovariances which lag each other by exactly zero days (C<sub>0</sub>). The reader may also note that C<sub>0</sub> is nothing more than the simple variance of all the observations. Thus, the autocorrelations for any lag from zero on up to N-1 may be calculated by the formulas (11) and (12). The display of these autocorrelations in a bar chart showing the computed values plotted against progressively higher lags forms

what is commonly referred to as the <u>autocorrelation</u> function.

Figure 2 is an example of such a chart. autocorrelation function along with a nearly identical illustration of the partial autocorrelation function (Figure 3), form the fundamental tools in the iterative procedure of identifying the model which fits a given time series. Whereas the shape of the autocorrelation function allows a guess as to the type of model which should be entertained (eg., AR, MA, ARMA, etc.) the partial autocorrelation function is used to help determine the order of the model. Box and Jenkins liken the use of the partial autocorrelation function to that of "deciding on the number of independent variables to be included in a multiple regression" (Box and Jenkins, 1976:64). They also include a detailed discussion of how to obtain the partial autocorrelation function (Box and Jenkins, 1976:64-65). Basically, the partial autocorrelation function shows the autocorrelation at higher lags after the effects of variables up to a certain lag have been regressed out. Therefore, the partial at lag 4 is the autocorrelation at lag 4 having accounted for the effects of lag 1, 2, and 3.

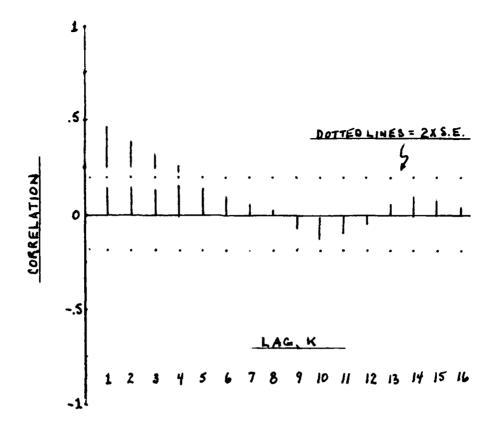


Fig 2. Example Autocorrelation Function

An example to clarify the above discussion is justified. Referring to Figure 2, let the assumption be that the autocorrelation function shown represents a time series which is under study. One can see that the locus of points formed by an imaginary curve connecting the tops of the charted values has a shape which is characteristic of some sort of exponential function.

According to Box and Jenkins, this is a classic

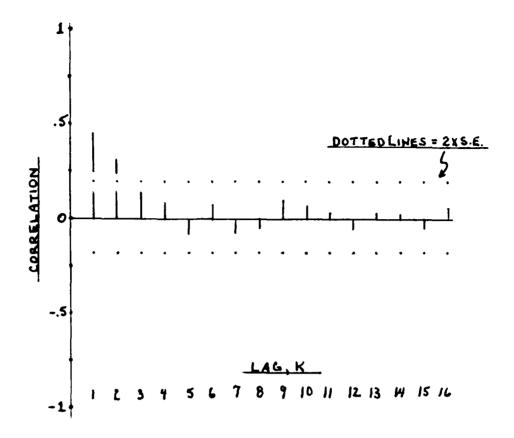


Fig 3. Example Partial Autocorrelation Function

indication of an autoregressive process. Notice also that the values of the autocorrelations drop off rapidly as the lag, k, increases. This is indicative of a stationary process as was discussed earlier in this chapter. Thus, the general type of model has been deduced. The problem remaining is to specify the order of the model.

One point should be made before determining the Box and Jenkins give formulas for calculating the standard error of the autocorrelation or partial autocorrelation at any chosen lag (Box and Jenkins, 1976:34.65). Twice the standard error is chosen as the level of significance for any calculated autocorrelation or partial autocorrelation value. In other words, if the computed correlation value for any given lag k does not exceed twice the standard error, then that correlation is effectively regarded as having a value of zero. Again referring to Figure 2 and 3, the dotted lines on each bar chart reflect twice the standard error at each lag. Observe that even though noticeable values have been computed past the second lag on the partial autocorrelation plot, in terms of the indicated standard error they are insignificant and thus are regarded as zero. In viewing the partial autocorrelation function (Figure 3), one can see that the function shows a definite cutoff after the lag k = 2 indicating that the process being analyzed is very likely a second order series. This being the case, the model would be initially estimated as AR(2) and would thus have the form

 $\tilde{z}_1 = \phi_1 \tilde{z}_{1-1} + \phi_2 \tilde{z}_{1-2} + a_1$ 

Similar procedures are followed for all time series. The shape and character of the autocorrelation and partial autocorrelation functions are thus the determining factors in the initial guess as to the model type. Box and Jenkins devote chapter 6 to the discussion of how to identify the type of model along with estimating its parameters ( $\phi_1$  and  $\phi_2$  in this case) (Box and Jenkins, 1976:171-207).

The reader may now be wondering how the parameters of the entertained model are calculated. As already stated, a significant amount of detail is contained in the text regarding model parameter estimation (Box and Jenkins, 1976:187-193,208-284). Let it suffice to say that a modern statistical software package such as BMDP allows the estimation of these parameters with a minimum of effort (Dixon and others, 1985:644-645). The main task of the analyst is to merely decide on the type and order of model he/she would like to fit to the data.

# Model Checking

After the model is identified and its parameters estimated, the model must then be tested to see whether a good fit has indeed occurred. If the fit is good, then the model may be retained for forecasting. If not,

then a new model must be entertained and likewise tested. This process continues until a good fit is obtained. Box and Jenkins refer to the testing of the model as "model diagnostic checking" (Box and Jenkins, 1976:285-299). One of two methods they describe to do diagnostics is that of displaying "the autocorrelation function of the residuals" (Box and Jenkins, 1976:285).

Earlier in this chapter, it was stated that the time series models of Box and Jenkins employ a white noise process (a series of normal shocks with mean zero and variance  $\sigma_{a^2}$ ). In so doing, this white noise is transformed into a model which adequately represents the series under study by linear filtering. The model thus derived can be rearranged so that estimated values of the originally input white noise process may then be computed. When this is done, the estimated white noise (or shocks) are designated as at and are also called the residuals (Box and Jenkins, 1976:289). The plot of the autocorrelation function of these residuals is one of the primary methods used to diagnose the adequacy of the model. If the autocorrelations of the residuals are significant at any lag, then a better model can most likely be found. The text suggests ways to use the residual autocorrelations to implement a better model (Box and Jenkins, 1976:298-299). If they are

insignificant, then the model is adequate and may be used for forecasting. Again, twice the standard error determines the significance of the computed autocorrelation at each lag. This method is easily accomplished with the BMDP package and used exclusively in this study.

To recap, the building of a time series model is an iterative process (Box and Jenkins, 1976:19). In most cases, a five step process takes place:

- (1) The general class of models is postulated. A plot of the actual time series may reveal the type of model to the "practiced eye".
- (2) The initial model to be entertained is then identified from the autocorrelation and partial autocorrelation functions.
- (3) The parameters of the entertained model are estimated.
- (4) The model is diagnostically checked by examining the autocorrelation of the residuals computed from the newly fitted model with its estimated parameters.
- (5) If the model's residuals are insignificant, the model may be used for forecasting. If not, a new model must be suggested based on the residual autocorrelation plot examination and steps 2 through 5 reaccomplished until a suitable model is found.

## Transfer Function Models

As discussed earlier, a <u>transfer function</u> links the output time series of a given dynamic system with the input time series to that system. From this perspective, one may be better able to understand the actual dynamics of the system in a purely mathematical sense if not in the physical sense. Such mathematical

insights are almost certain to eventually lead to better physical understanding as well.

Box and Jenkins go into considerable detail on the subject of transfer function modeling in chapters 10 and 11 of their text (Box and Jenkins, 1976:335-420). In perhaps their best illustration of how a transfer function model is formed, they exhibit the necessary steps in an example problem concerning the output of carbon dioxide (CO2) from a gas furnace which has influential inputs of air and methane (CH4) (Box and Jenkins, 1976:381-386). In a very fortunate coincidence, the BMDP manual, in discussing their own example problem to illustrate transfer function modeling, uses the same gas furnace problem of Box and Jenkins (Dixon and others, 1985:650-655). reader is able to match the logic used in forming the computer solution to the problem with the theoretical logic contained in Box and Jenkins. For this reason, further discussion of transfer function modeling will be limited to the main tool of analysis in this method which is the cross correlation function. The reader desiring more detail on transfer function modeling is therefore urged to obtain a copy of Box and Jenkins and the BMDP manual.

Just as the autocorrelation and partial autocorrelation functions yield much information about a given single time series, the cross correlation function reveals much about the relation between any two time series. A cross correlation between two time series may be appropriate at any time one believes that the two interact in some sort of a dynamic system. As brought out in the literature review, this is the belief concerning the energetic electron flux levels and the solar wind speed/IMF B2. Formulae for determining the cross correlation at any desired lag between any two dynamically related time series are contained in the text, and the reader will notice that they are very similar to those given for determining the autocorrelations (Box and Jenkins, 1976:374). formulae are as follows:

$$r_{xy}(k) = c_{xy}(k)/s_x s_y \tag{13}$$

where

$$c_{xy}(k) = -\sum_{t=1}^{n-k} (x_t - \overline{x})(y_{t+k} - \overline{y}) \quad k = 0, 1, 2, \dots$$
 (14)

or

$$C_{xy}(k) = -\sum_{t=1}^{n+k} (y_t - \overline{y})(x_{t-k} - \overline{x}) \quad k = 0, -1, -2, \dots$$
 (15)

In equation (13),  $r_{xy}(k)$  is the cross correlation at lag k between an input series x and an output series y. The term  $c_{xy}$  is the cross covariance between the two series, and equations for obtaining it are shown in (14) and (15) where n stands for the number of pairs of series values  $(x_t, y_t)$  and  $\overline{x}$  and  $\overline{y}$  are the respective series means.

Once the cross correlations at each lag are determined, they are plotted in a bar chart which is very similar to Figures 2 and 3 shown earlier in this chapter. This plot is referred to as the cross correlation function. From this chart the model is tentatively identified. An estimation of the dead-time parameter, b, may be made directly from the cross correlation function along with estimates of the parameters r and s from the difference equation (8). The parameter b is usually set equal to the value for the first positive lag at which a significant cross correlation occurs. Thus, if the first significant cross correlation for two dynamically connected series occurs at lag k=2, then b is set equal to 2. The model estimation and final identification process then proceeds iteratively as it did for the single series models.

One final point which should be made is that the method of transfer function modeling used by BMDP2T is that of "prewhitening the input" (Box and Jenkins, 1976:379-380). This method essentially recognizes that two series which are dynamically related also have an element of noise involved. Thus, any output series modeled from a related input series is influenced not only by the input but by noise as well. The amount of noise present in the model can thus be an indicator of how deeply related any two series are. Said another way, the greater the noise component, the less influence the input has on the output series. This noise component is modeled in the transfer function models determined in this study. Also, and perhaps more importantly, the prewhitening method is used since it can provide initial rough estimates of the  $v_k$  parameters of the transfer function equation (10) (Box and Jenkins, 1976:380). With these initial rough estimates, the model fitting procedure can usually be implemented in a more efficient manner.

# BMDP Examples

Appendix E of this report contains three example time series analysis problems as performed using program

BMDP2T. Example 1 was done on the SSC, while Example 2 and Example 3 were done on the CYBER.

Example 1 is typical of the early stages of identification of a model for any given time series. this particular run, energetic electron channel SEEI is analyzed, and the plot of the autocorrelation and partial autocorrelation functions indicate a stationary series with autoregressive tendencies. In explanation, the autocorrelation function decreases rather quickly as the lag, k, increases, while the partial autocorrelation function has a definite cutoff after lag k = 2 (page 138). This would indicate a probable best fit of an AR model with first and second order terms. However, the analyst, not being seasoned in the "art" of time series analysis, decided to attempt to fit an AR model with first and 27th order terms as shown in the ARIMA paragraph on page 139 of Appendix E. On the page 140, the results of the attempted fit yield a model with estimated parameters of 0.7856 and 0.1039 (by the backcasting method). The reader will notice that the T-RATIO of the first order is very high (23.16) indicating that the model should most definitely contain a first order AR term. However, the T-RATIO of the 27th order is only 1.91 indicating that perhaps this parameter is inappropriate. Nevertheless, if one

believes in the validity of the parameters as estimated, the model would be

 $(1 - 0.7856B - 0.1039B^{27})$  $\tilde{z}_{t} = a_{t}$ 

where  $\mathbf{Z}_{t}$  represents the current estimated deviation of SEEI from the series mean. Using the plot of the autocorrelations of the residuals,  $\mathbf{a}_{t}$ , resulting from the attempted fit of the model (page 141), the reader can see that the fit is inappropriate since the residuals show significant (ie., greater than two standard errors) autocorrelations at lags k = 1, 2, and 4. Thus, the analyst should try to fit a model which better explains the data. The best guess of the author would be, as stated earlier, an AR model with first and second order terms. This iterative identification process is an illustration of the "art" of time series analysis. The person with a better of grasp of this kind of analysis will be able to converge to a more appropriate model in a fewer number of tries.

Example 2 (page 144) is much the same as Example 1. Here the series under study is energetic electron channel SEEIII. The analyst has already examined the autocorrelation and partial autocorrelation functions of the series and thus has decided to spare computer processing time by not including the autocorrelation

functions in the analysis. In the ARIMA paragraph on page 145, an attempt is made to fit an ARMA model with AR terms of first and 27th order along with an MA term of first order. Using the parameters estimated by the backcasting method, the model is

 $(1 - 0.9850B - 0.1701B^{27})^{2}_{i} = (1 + 0.1227B)a_{i}$ 

The fit of this model is somewhat better than that of Example 1 as evidenced by the lesser number and magnitude of the significant autocorrelations of the residuals (page 147). The analyst may want to accept the fitted model or reiterate and attempt to fit a different one. A forecast of future values is also made (page 148) based upon the model as estimated above. The reader can see that the standard error of the forecast is 0.02037 for the initial day. It should also be noted that the standard error increases as one attempts to forecast farther into the future. The analyst may also use the forecast paragraph to check the accuracy of the fitted model by requesting that the forecasting take place over a period for which actual series values are available. In the example, days 345 through 365 have actual values which may be compared to the forecast values.

Example 3 is a transfer function model visualizing the IMF Bz component as the input series and the SEESSD channel as the resulting output. important point to note here is that the plot of the cross correlations (page 153) shows only one significant value at lag k = 1. Thus, the analyst might surmise a very simple transfer function model with the delay parameter, b, set equal to 1 and parameters r and s (the left side and right side orders of the difference equation (8)) both equal to zero. Hence, (r,s,b) would be (0,0,1). The analyst would then use the procedure described in Chapter 10 of Box and Jenkins to derive initial estimates of of the operators  $\delta(B)$  and  $\omega(B)$ (Box and Jenkins, 1976:345-351). With these initial estimates, BMDP2T is then run interactively in a series of trial and error steps until a suitable model is identified (Dixon and others, 1985:650-656). iterative part of developing a transfer function model is mainly contained in deriving a good model for the input series and for the noise component. Also, as discussed in Box and Jenkins, there may be more than one transfer function model. That is to say, r, s, and b may be set equal to other values and other models may be fit with equally good (or poor) results (Box and Jenkins, 1976:387). The final transfer function model

derived here and used to obtain the forecasts on page 161 is

 $Y_t = -24.01BX_t + (1/(1 - 0.6125B - 0.1973B^{27}))a_t$ 

The terms in parentheses represent the noise portion of the model. The first term on the right hand side represents that portion of the output series (where  $Y_t$  is the output or SEESSD in this case) which may be explained directly by the input series (where  $X_t$  is the input series or IMF  $B_z$ ).

Again, the interested reader is strongly advised to obtain a copy of both the text and the BMDP manual in order to fully grasp the time series analysis procedures. The results of the Box and Jenkins method as applied to the data received will now be presented.

### V. Results

# Individual Channel Models

Tables I - IV beginning on the next page summarize the time series analysis results for each of the energetic electron channels. In other words, the series of discrete daily average values from each channel (SEESSD, SEEI, SEEII, and SEEIII) were separately analyzed in order to try to fit an appropriate model to each individual channel.

The first thing the reader will notice is that no individual models for SEEIV are included. The reason for this is simply that the data in the SEEIV channel did not lend itself to this type of analysis. SEEIV contains recorded values of energetic electron count rates in the range of 9.7 - 16 MeV. An initial examination of the SEEIV data over the last 765 days (from April 84 until May 86) revealed that the autocorrelations were effectively zero at all lags save one: At lag k = 35, the computed autocorrelation was 0.199 which was significant. Likewise, the computed partial autocorrelations were all effectively zero except at lag k = 35 where the computed value was once again 0.199. The fact that nothing could be said about the SEEIV data due to the lack of any identifiable

TABLE I

Suggested Models for the SEESSD Channel (1.2 - 1.8 MeV)

Data Length	;	Suggested Model	S.E.
100 days (Jan 86-May		.8250B1776B <sup>27</sup> )z <sub>t</sub> = (1 + .4401B)a <sub>t</sub>	
365 days (May 85-May	86)	$.7470B2329B^{27})z_{t}$ = $(1 + .2727B)a_{t}$	•

Approximate series mean = 224.49

pattern in the autocorrelations meant that this data, by itself, could not be analyzed via time series analysis. However, transfer function modeling of this channel was possible and will be shown later. The reader will recall that the autocorrelation functions are the primary tool in identifying a series model. With no values other than zero, identification becomes impossible. Sample autocorrelation and partial autocorrelation functions for all the energetic electron channels are included in Appendix F.

In examining Table I, the reader will notice that a couple of different models are identified. The first model is an AR1,27 MA1 representation of the SEESSD channel over the last 100 days of the available data (27 Jan 86 - 6 May 86). The second model is a

similar representation of the last year of the data (7 May 85 - 6 May 86). These two models represent the "best fits" that the author was able to identify in terms of (1) the least residual mean square values and (2) the smallest forecast standard errors. In both cases, the plot of the autocorrelation function of the residuals showed a good fit (no significant correlations at any lag). Many other models were fitted to the data with varying degrees of success which were all less than that achieved by the two models listed. The column entitled "S.E." represents the forecast standard error for the given model. In Table I, two such values are shown per model. The reason for this is to illustrate that a given model can achieve varying amounts of effectiveness at predicting the series values depending upon where the initial point of forecast takes place. In the case of the 100 day model, two different requests for forecasting (predictions) were made. The one which resulted in a S.E. of 157.49 began at the 80th day of the 100 days of data while the other which yielded an S.E. of 174.29 began at the 71st day. The forecasts were requested beginning at these particular days in order to allow the available SEESSD series values to be compared to the values predicted by the models. Similarly, the 365 day model resulted in an S.E. of

227.34 and 241.04 for requested forecasts beginning at the 350th and 200th days of the data respectively. The approximate series mean for each of the channels is shown at the bottom of each of Tables I - IV. This is shown in order to give the reader an idea of the magnitude of the forecast standard error as compared to the mean series value. These means are only approximate since they were initially determined over the entire 4 years of data originally presented for analysis by LANL. Thus, the mean of each series will change depending on the amount of data (number of days) to which a model is fit. It was assumed here that the means of the lesser series do not vary substantially from those determined for the entire four years of the data. This is a reasonable assumption since each of the series for SEESSD, SEEI, SEEII, and SEEIII appear to be stationary (ie., the mean does not vary considerably over time).

The models shown in Tables II - IV are displayed in a similar manner to those in Table I and thus represent the best models for the SEEI, SEEII, and SEEIII channels. As in the case of the SFESSD channel, each of the models for the other three channels appears to be best represented by some type of ARMA model. Indeed, a closer look reveals that almost without exception each of the models for the different channels

contains autoregressive terms of first and 27th order along with a moving average term of first order. In each of the tables, the term  $z_{\,t}$  should be taken to

TABLE II

Suggested N	Models for the SEEI Channel (3.4	- 4.9 MeV)
Data Length	Suggested Model	S.E.
100 days (Jan 86-May	(17975B0973B260421B27)zt $= (1 + .0959B)at$	.336
365 days (May 85-May	(16243B0806B260844B27)zt $= (1 + .4623B)at$	1.598
	Approximate series mean = 1.05	

TABLE III

Suggested N	Models for t	he SI	EII Channel (	4.9 -	6.6 MeV)
Data Leng	gth	Sugg	ested Model		3.E.
100 days (Jan 86-May	86)	(1 -	.9240B)zt = (12931B	i)a <sub>t</sub>	.1195
365 days (May 85-May	86)	(1 -	.7043B)z <sub>t</sub> = (17043B	s)a <sub>t</sub>	.1203
	Approximate	e seri	es mean = .14	58	

TABLE IV

Suggested Models	for	the	SEEIII	Channel	(6.8)	; –	9.7	MeV)	
------------------	-----	-----	--------	---------	-------	-----	-----	------	--

Data Leng	th Suggested Model	<u>S.E.</u>
100 days (Jan 86-May	(19932B1222B27)zt = (1 + .1307B)a <sub>t</sub>	.0167
365 days (May 85-May	(19850B1701B27)zt $= (1 + .1227B)at$	.0204
	Approximate series mean = 14.01	

mean the current value of the stated energetic electron channel as it deviates from the overall series mean. This term was explained in the previous chapter.

# Transfer Function Models

The results of the transfer function modeling are presented in Tables V - XVII. Tables V - VIII contain models suggested for each of the energetic electron channels using the entire extent of the available cross correlated data (706 days or from 8 May 83 to 12 Apr 85). The models include those for which the solar wind as well as the IMF B<sub>2</sub> component is considered as the input series. The output series is always considered to be the stated channel. Thus, in each of the suggested models in those tables, the term y<sub>1</sub> denotes the output series current value (the channel) and the term x<sub>1</sub> denotes the input series (either the B<sub>2</sub>

component or the solar wind). The term associated with  $a_t$  represents the noise component in the model. Tables IX - XII are duplicates of V - VIII except that the models shown only represent data covering the last 200 days of cross correlation (25 Sep 84 - 12 Apr 85). Tables XIII - XVI are once again the same except that they only represent some 70 days worth of cross correlated data. Tables XVII and XVIII are special in that they identify transfer function models for the SEEIV channel based on a solar wind input. As stated earlier in this chapter, an individual series model for the SEEIV channel was not identifiable. Each of the tables exhibits the r, s, and b values used in the original difference equation along with the standard error of the forecasts. The r, s, and b values were explained in Chapter 4. As is evident from the tables, many of the models contain delay parameters equal to one day, though some have delays which differ significantly from one.

Table V may be used in illustration. The models for the  $B_z$  and solar wind inputs are as shown. The r, s, and b used initially were (0,0,1) and (1,2,1) respectively. The forecast standard errors for the models shown are 186.91 and 187.36.

### TABLE V

Transfer Function Models for the SEESSD Channel 706 days (8 May 83 - 12 Apr 85)

# 

### TABLE VI

Transfer Function Models for the SEEI Channel 706 days (8 May 83 - 12 Apr 85)

```
Input Suggested Model

Bz  yt = -.083Bx_t + (1/(1 - .5589B - .1431B^3 - .1091B^2 - .2387B^2 ))a_t

(r,s,b) = (0,0,1)
Solar

Wind  yt = ((.0020B^2 - .0013B^4)/(1 + .9998B))x_t
+ (1/(1 - .5680B - .0444B^2 - .1951B^3 + .1515B^4 - .1906B^2 ))a_t

(r,s,b) = (1,3,1)
S.E. = 1.497
```

# TABLE VII

Transfer Function Models for the SEEII Channel 706 days (8 May 83 - 12 Apr 85)

# Input

# Suggested Model

```
B<sub>2</sub> y_t = -.0088Bx_t + (1/(1 - .6468B - .1165B^3 - .1950B^27))a_t
(r,s,b) = (0,0,1)
S.E. = .1699
Solar
Wind y_t = .0002B^2x_t + (1/(1 - .6974B + .1173B^2 - .1039B^3 - .1604B^27))a_t
(r,s,b) = (0,0,2)
S.E. = .1689
```

### TABLE VIII

Transfer Function Models for the SEEIII Channel 706 days (8 May 83 - 12 Apr 85)

### Input

### Suggested Model

```
B_{2} \qquad y_{t} = (-.0713B^{5} + .1351B^{6})x_{t} \\ + (1/(1 - .3434B - .1771B^{2} - .1551B^{3} + .1935B^{4}))a_{t} \\ \qquad (r,s,b) = (1,1,5) \\ s.E. = 2.058
Solar
Wind y_{t} = ((-.0023B^{7} - .0023B^{8})/(1 - .6482B))x_{t} \\ + (1/(1 - .3343B - .1813B^{2} - .1504B^{3} + .1911B^{4}))a_{t} \\ \qquad (r,s,b) = (1,1,7) \\ s.E. = 2.196
```

# TABLE IX

Transfer Function Models for the SEESSD Channel 200 days (25 Sep 84 - 12 Apr 85)

# 

# TABLE X

Transfer Function Models for the SEEI Channel 200 days (25 Sep 84 - 12 Apr 85)

```
Input Suggested Model

B<sub>2</sub>  y<sub>1</sub> = -.2078Bx<sub>1</sub> + (1/(1 - .4547B - .0881B<sup>3</sup> - .0876B<sup>2</sup>1 - .3046B<sup>2</sup>7))a<sub>1</sub>

(r,s,b) = (1,2,1)
S.E. = 1.556

Solar Wind y<sub>1</sub> = ((.0026B - .0047B<sup>3</sup>)/(1 - .9334B))x<sub>1</sub>
+ (1/(1 - .4937B - .2745B<sup>3</sup> +.2193B<sup>4</sup> - .1864B<sup>2</sup>7)a<sub>1</sub>

(r,s,b) = (1,3,1)
S.E. = 1.546
```

#### TABLE XI

Transfer Function Models for the SEEII Channel 200 days (25 Sep 84 - 12 Apr 85)

# Input Suggested Model

 $B_z$   $y_t = -.0115Bx_t + (1/(1 - .5718B - .3042B^3 + .2504B^4 - .3065B^2^6))a_t$ 

(r,s,b) = (0,0,1)S.E. = .1889

Solar Wind  $y_t = .0006B^2x_t + (1/(1 - .5061B - .2915B^3 + .1598B^4 - .3445B^2))a_t$ 

(r,s,b) = (0,0,2)S.E. = .1810

### TABLE XII

Transfer Function Models for the SEEIII Channel 200 days (25 Sep 84 - 12 Apr 85)

# Input

# Suggested Model

B<sub>z</sub> Parameter estimation terminated, no apparent model.

Solar Wind  $y_t = .0022B^9x_t + (1/(1 - .8535B + .1603B^2 - .0439B^6 + .3125B^{13} - .4784B^{14}))a_t$ 

(r,s,b) = (0,0,9)S.E. = 1.221

### TABLE XIII

Transfer Function Models for the SEESSD Channel
Bz = 70 days (30 Oct 83 - 7 Jan 84)
Solar Wind = 71 days (1 Feb 85 - 12 Apr 85)

# Input

# Suggested Model

B<sub>2</sub> No significant cross correlations, no apparent model.

Solar

Wind  $y_t = .2161B^3x_t + (1/(1 - .7991B))a_t$ 

(r,s,b) = (0,0,3)S.E. = 163.42

#### TABLE XIV

Transfer Function Models for the SEEI Channel  $B_z = 70$  days (30 Oct 83 - 7 Jan 84) Solar Wind = 71 days (1 Feb 85 - 12 Apr 85)

# Input

# Suggested Model

 $B_z$  No significant cross correlations, no apparent model.

Solar

Wind  $y_t = .0005B^2x_t + (1/(1 - .7357B + .2114B^2))a_t$ 

(r,s,b) = (0,0,2)S.E. = 1.218

# TABLE XV

Transfer Function Models for the SEEII Channel  $B_z = 70$  days (30 Oct 83 - 7 Jan 84) Solar Wind = 71 days (1 Feb 85 - 12 Apr 85)

### Input

### Suggested Model

```
B_{z} \quad y_{t} = (-.0191B^{11} + .0453B^{12})x_{t} + (1/(.6368B - .1913B^{4}))a_{t} + (1/(.6368B - .1913B^{4}))a_{t} + (1/(.6368B - .1913B^{4}))a_{t}
S_{z} = (0,1,11)
S_{z} = .1244
Solar
Wind
y_{t} = ((.0007B^{3} - .0010B^{6})/(1 + .3376B))x_{t} + (1/(1 - .5547B))a_{t} + (1/(1 - .5547B))a_{t}
(r,s,b) = (1,4,2)
S_{z} = .1446
```

# TABLE XVI

Transfer Function Models for the SEEIII Channel  $B_z = 70$  days (30 Oct 83 - 7 Jan 84) Solar Wind = 71 days (1 Feb 85 - 12 Apr 85)

# <u>Input</u>

# Suggested Model

B<sub>z</sub> No significant cross correlations, no apparent model.

Solar Wind  $y_t = (.00009B^2 + .00005B^3 + .00003B^4)x_t + (1/(1 - .3589B + .2715B^2 - .4446B^25 - .4600B^27))a_t$  (r,s,b) = (0,2,2) S.E. = .0136

# TABLE XVII

Transfer Function Models for the SEEIV Channel 200 days (25 Sep 84 + 12 Apr 85)

#### Input

# Suggested Model

Βz

No significant cross correlations, no apparent model.

Solar

Wind

 $y_t = (.00016B^9 - .000012B^{10})x_t + (1/(1 - .7571B + .0728B^2 + .0810B^5 - .1919B^{15} + .0701B^{20}))a_t$ 

(r,s,b) = (0,1,9)S.E. = .0932

#### TABLE XVIII

Transfer Function Models for the SEEIV Channel  $B_z = 70$  days (30 Oct 83 - 7 Jan 84) Solar Wind = 71 days (1 Feb 85 - 12 Apr 85)

# Input

### Suggested Model

 $B_z$  No significant cross correlations, no apparent model.

Solar

Wind  $y_t = (.00005B^2 - .00003B^4)x_t + (1/(1 - .7078B - .2644B^27)a_t$ 

(r,s,b) = 0,2,2)S.E. = .01846 Additional data concerning the autocorrelation and partial autocorrelation functions of the two input series (the IMF B<sub>2</sub> and the solar wind speed) are contained in Appendix G. These ACFs and PACFs were used to identify the input series necessary for the initial part of the transfer function modeling as accomplished by the prewhitening method. Both of these input series were shown to be AR(1) processes. Also, Appendix H is a listing of the 70 day cross correlation functions between these input series and the various energetic electron channels (output series) so that the reader may better grasp the procedures for identifying the r, s, and b values in transfer function modeling.

# Discussion

Apart from the time series models developed in this study, good support for some of the results obtained in prior studies of energetic electron flux is evident. Perhaps the two best examples of this are (1) the fact that a large positive autocorrelation showed up very consistently near lag k = 27 in the ACFs of the lowest three energetic electron channels studied (SEESSD, SEEI, SEEII), and (2) the delay factor for many of the transfer function models was on the order of one to two days. Finding (1) above is in good agreement

with McCormick who stated in his conclusions that "whatever processes affect electron fluctuations in the three lowest channels are different from those affecting the two highest channels" (McCormick, 1984:62,63). addition, it supports the findings of Paulikas and Blake who noted the variability of energetic electron fluxes associated with the 27 day solar rotation period Paulikas and Blake, 1978:22). Finding (2) is also in harmony with McCormick's finding that the electron fluxes in the SEESSD, SEEI, and SEEII channels showed a weak but evident correlation with one and a half to two-day old solar wind speed data. (McCormick, 1984:62). Moreover, Paulikas and Blake are in agreement with this delay factor (Paulikas and Blake, 1978:11). Most of these previous findings were mentioned in Chapter II of this report.

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Referring to the individual series models,

Tables I - IV seem to suggest mixed amounts of success in using time series analysis. With regard to Table I, the best models identified yielded a forecast standard error which is undoubtedly too great to make these models of any practical use. A 95% confidence interval for a prediction one day in the future using either of the two models listed would be an extremely broad interval of about 3 to 4 times the mean value of the

series itself. One should bear in mind that the standard error of the forecasts depends somewhat upon where (ie., beginning on what day) the requested forecast is made as is illustrated by the two different standard errors listed for each model in Table I.

Still, this appears to render the models unacceptable.

A more reasonable model would give a 95% CI of, say, no more than half of the mean value of the series. With this in mind, it then appears that the models in Tables II and III are likewise of little value.

The results shown in Table IV, however, offer some hope. As shown, the models for the SEEIII channel both yielded very small forecast standard errors (.0167 and .0204 for the 100 day and the 365 day models respectively) as compared to their respective series This would appear to make them of possible future use, but one must keep in mind the data as it appeared in this channel. The SEEIII values were characterized by fairly constant count rates (usually about 0.12 - 0.15) except on a few rare occasions where the count rate skyrocketed to values such as 1417.89 which was the highest value recorded in this series. Outliers such as this can have a marked effect on the series mean and standard deviation and thus make a given model seem more appropriate than it may in fact be. In

truth, with the exception of the few outliers in the SEEIII channel, one might very well be able to forecast the next two days of count rates for SEEIII simply by "eyeballing" the past data and noting that, for the most part, it varies little from day to day.

Many of the transfer function models are likewise of doubtful utility. In fact, it seems that the standard error for the forecasts based on any given transfer function model was often times as much or greater than that of the individual series models.

Again, bearing in mind the series means and the forecast standard errors indicated, it appears that the 706 day models shown in Tables V, VI, and VII are not useful.

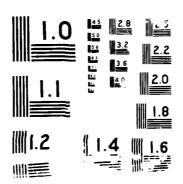
Once again though, the models for the SEEIII channel (Table VIII) have acceptable standard errors as compared to the series mean and thus may be of some possible value in prediction.

The exact same set of statements may be said in regard to the 200 day models (Tables IX - XII). Once again the 200 day model for the SEEIII channel (based on a solar wind input) possessed a small standard error as compared to the mean value of the SEEIII channel. The reader will note that a transfer function model for the SEEIII channel using the IMF B<sub>2</sub> as the input was not identifiable. The computer terminated parameter

estimation due to estimates of the parameters "becoming too highly correlated".

In recognition of the fact that quite a bit of the input series data had to be estimated (ie., the data from the NSSDC on the solar wind and IMF B2 contained many missing values), it was felt that perhaps the transfer function models were not getting a fair chance at success. Thus, a further examination of the data was done in an effort to determine a "stretch" of erough consecutive days of unestimated data so as to make the transfer function models less susceptible to any bias caused as a result of all the estimated values. For the IMF B2 data, the longest stretch of "clean" data which could be cross correlated with the energetic electron channels occurred during the dates from 30 Oct 83 to 7 Jan 84 (70 days). For the solar wind, the longest stretch occurred between 1 Feb 85 to 12 Apr 85 (71 days). The results of the transfer function models using these days' B2 and solar wind values are given in Tables XIII - XVI. As is evident, results using this "unbiased data" were only marginally better than those when the data containing estimated values were used. And, once again, the transfer function model for the SEEIII channel offered the most possible usefulness. The reader may note that in three cases, models were not

R TIME SERIES AWALYSIS OF ENERGETIC ELECTRON FLUXES (12 - 16 MEY) AT GEOS (U) AIR FORCE INST OF TECH MERIGHT-PATTERSON AFB OH SCHOOL OF ENGI M P HALPEN DEC 86 AFIT/GSO/ENS-ENP/860-1 F/G 22/1 HD-R194 368 UNCLASSIFIED NL



identifiable due to lack of any significant cross correlations between the input series and the stated energetic electron channel. In all three cases, this occurred when the  $B_z$  component was used as the input series. This again offers some support for the findings of McCormick who noted that little correlation existed between the IMF  $B_z$  and the energetic electron fluxes (McCormick, 1984:63).

Tables XVII and XVIII represent the only identified way of modeling the SEEIV channel. In both cases, no model could be identified using the IMF B<sub>2</sub> as the input. However, when the solar wind was assumed as the input, some models of possible usefulness were obtained as the standard error of the forecasts was small when compared to the series mean. Once again, the same things which were said about the behavior of the SEEIII channel may also be said about the SEEIV channel: The series contains outliers. Therefore, careful use of these models is in order.

### VI. Conclusions and Recommendations

# Conclusions

This study had some measurable "positives" to report. First, it reinforced previous findings that changes in the levels of energetic electron flux lag solar wind speed changes by one to two days. brought out by the fact that many of the suggested transfer function models had delay parameters (b) of 1 or 2 which indicates that the output (the given electron channel) lags the input (the solar wind or the IMF B2) by a one to two day time period. Second, the fact that the autocorrelation functions of the three lowest energy channels (in addition to that of the solar wind) showed a strong positive autocorrelation around the lag k = 27definitely supports previous findings that the solar rotation cycle plays a considerable part in the makeup of the solar wind plasma and, consequently, in the changes in energetic electron flux. A closer inspection of the ACFs shows that a buildup of autocorrelations at lags just before the k = 27 day point occurs followed generally by a "builddown" in the autocorrelations after this point. A possible theory on the reason for this is the surmised existence of a "hot spot" of the sun which emits greater than average amounts of plasma and which

slowly moves across the solar surface while the sun rotates. Third, the fact that the three lowest energy channels displayed this 27 day lag trend while the higher channels did not gives support to previous findings that the lower energetic electron channels appear to behave differently from the higher energy channels. Thus, the control mechanisms for the lower energy channels seem to differ from those of the higher channels.

Of less positive note, it seems that the entire methodology of time series analysis did not apply as well as had been hoped to the data presented for study. This might lead one to become disconsolate with regard to the Box and Jenkins method. However, some qualifications which affected the analysis have to be emphasized.

First, it should be recognized that the data were anything but consistent. The LANL data on the energetic electron count rates were by far the best with regard to consistency. Of the data from LANL actually used, only 15 days worth of missing values had to be estimated. However, this was far less than the data from the NSSDC on the solar wind and IMF Bz which contained stretches of days at a time where no values

were posted. The fact that data were missing at all can only tend to have adverse effects on the analysis.

Second, there is the iterative or "artful" nature of time series analysis which must be considered. The Box and Jenkins method, as presented in BMDP, requires that the user interact with the program in what is frequently a very long and time consuming set of processes before a model is finally identified. user must master the art of examining the autocorrelation and partial autocorrelation functions to unmask the true behavior of a given time series. Obviously, the more experience one has at this sort of thing, the quicker the process of identifying the appropriate model becomes. The same may be said more emphatically of transfer function modeling. Mastery of the secrets held in the cross correlation function between a given input and output series is of the utmost importance. Without it, an attempt to derive a transfer function model may be doomed from the start. In the transfer function modeling process as presented in BMDP, the user is required many trial and error interactions. Initially, assuming significant cross correlations exist between the two series, a guess at r, s, and b may be made and the appropriate estimates of the difference equation parameters (all figured by hand) must be input

into the computer. The computer will further refine these estimates or else "throw them out entirely". When this occurs, the user is forced to make a new guess at r, s, and b and refigure new parameter estimates. Also, a great deal of the time spent in transfer function modeling consists of trying to iteratively determine the noise portion of the model (ie., the part attached to In addition, one is reminded in Box and Jenkins that for transfer function modeling, there is not necessarily one unique model (Box and Jenkins, 1976:387). Therefore, despite the fact that the model one may derive seems poor (or even good), the analysis, if carried out a different way, might well yield more favorable results. Different surmisals of the r, s, and b parameters can, for instance, lead to different models. In short, there is often too much left to the skills (and whims) of the user and not enough left to the computer, at least as far as the BMDP implementation is concerned. Until a more precise way of performing a time series analysis is developed so as to remove a lot of the "art", this method may well be scorned by many when, in fact, it may have the potential of modeling a given equispaced series of values very accurately.

Third, one must consider the physical processes at work in the magnetosphere. As has already been

the solar wind, the IMF, and the energetic electron fluxes are anything but trivial. Therefore in this respect, the results of this study are definitely in support of this non-simplistic theory, since most of the transfer functions derived here seem to indicate that the fluxes depend quite heavily on something other than just the solar wind and the IMF B<sub>2</sub>. If the solar wind and the B<sub>2</sub> component were the sole drivers of the energetic electron fluxes, then one would suspect that the forecast standard errors in all the transfer function models would, on the whole, be considerably less than what they turned out to be. Put simply, there is more at work in the magnetosphere with regard to energetic electron flux than is reflected here.

# Recommendations

Notwithstanding the somewhat disappointing results of the model fitting attempted in this study, the author continues to believe in the Box and Jenkins method and also that it has utility with regard to the further study of energetic electron fluxes and the magnetosphere in general. If nothing else, even with data that contained many missing values, the various time series studied here indicate a good capability to

be modeled by the Box and Jenkins method. supported by the fact that so many of the autocorrelation and partial autocorrelation functions exhibit what appear to be "classic" time series model Moreover, support is offered by the models tendencies. themselves. For instance, an examination of BMDP Example #3 in Appendix E shows that the forecasts made by the model are at least attempting to follow the trends of the actual values (page 162). Thus, it is recommended that time series analyses of such data continue to be carried out in the future. One possibility for shortening the long iterations involved in the BMDP implementation of the method is the development and use of more sophisticated software which will eliminate some or all of the guessing involved on the part of the user and make the method more the application of a precise set of steps which yields a very accurate model. From consultations with advisors and peers, it is the understanding that such computer software may be on the horizon. If so, it would definitely be to the advantage of the person doing a time series analysis to use such programs.

With regard to future data, it would be most helpful if it could somehow be made more consistent and required less by way of estimation of missing values.

This is truly an ideal, however, and the author realizes that it is almost as difficult to predict when a satellite sensor is about to malfunction as it is to predict electron fluxes. If "cleaner" data <u>could</u> be obtained, the possibility of doing a time series analysis based on hourly readings might be an interesting undertaking if the improved software spoken of above could also be used. If such a computer program were to yield an extremely accurate model for, say, a transfer function between the solar wind and the SEEIV channel <u>utilizing hourly data</u>, it might be more plausible to ask for (and rely on) a prediction 24 - 48 hours (time periods) in the future. The reader will recall that usually such predictions carry with them a very high standard error.

Additionally, nothing can help the potential user of the Box and Jenkins method more than a solid foundation in the material as presented in the text. As a minimum, a prior graduate level course in time series analysis is recommended.

Finally, with regard to the models derived here, only those with an acceptably small forecast standard error should be considered for use. The reader is reminded that time series modeling is not a static process in that as new data is recorded, the model should be updated to reflect the series' latest trends.



Appendix A: FORTRAN Program to Read the LANL Data



## Appendix A: FORTRAN Program to Read the LANL Data

program readom

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c c This program was written so that the dath on the Omni Tape could be compressed into a more manageable file of daily averages. The really important variables are the variables for the solar wind velocity (v) and those having to do with the IMF (bxe,bye,bze,bym,bzm). Note that some variables have been declared as integers and some as real depending on how the NSSDC specified their individual formats on the Omni Tape (eg., I2 or F6.2, etc.). Note also that the last 20 real variables beginning with the variable aptsi are the averages of their respective variables which we are interested in (eg., aptsi is the average of ptsi, av is the average solar wind velocity, etc.).

c

integer flag,yr,day,hr,brn,idimf,idsw,ptsi,ptsp,
\*kp,c9,r,dst,nn(20),dread

real b,f,thb,phb,bxe,bye,bze,bym,bzm,sigb,sigf,
\*sigbx,sigby,sigbz,t,n,v,phv,thv,sigt,sign,sigv,
\*sigphv,sigthv,
\*ss(20),
\*aptsi,aptsp,akp,ac9,ar,adst,ab,af,athb,aphb,
\*abxe,abye,abze,abym,abzm,at,an,av,aphv,athv

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The character variable junk was introduced as a variable into which any miscellaneous data records which do not contain numerical characters could be read and discarded. For the data set given, this occured only once: on the 10th hour of the 338th day of 1982, a row of asterisks was present. Thus, this data were discarded by having it read into the variable junk.

c

character junk

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 $\mathbf{C}$ 

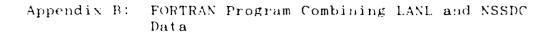
The general purpose directory tmp contained the raw Omni Tape data (hourly readings amounting to a 5 megabyte file). The file name containing the raw data was file3. Consequently, we must open this file in order to manipulate the data. Also, the file swdat must be opened in order to read in the newly averaged data.

```
The file swdat will be the compressed and more
      manageable file.
Ċ,
\mathbf{C}
      open(2, file='/tmp/file3')
      open(3, file='swdat')
C
      The following are small routines to set the
C
      summing variables back
      to zero at the beginning of a new day.
5
      do 6
              kk = 1,20
           ss(kk) = 0.0
6
      continue
              kk = 1,20
      du 8
           nn(kk) = 1
8
      continue
^{\circ}
      The variable dread is a conditional termination
C
      variable for the program.
C
\mathbf{C}
       dread = 0
10
       if(dread.eq.1) goto 200
C
      The following are the format statement and read
C.
       statement that are used to read the raw data from
C
       the file file3.
C
C.
       format(i1,i2,i3,i2,i4,2i2,2i3,14f6.2,f8.0,
11
      *f5.1,3f6.1,f8.0,f5.1,3f6.1,i2,i1,i4,i5)
C
       Note that the original raw data on the Omni Tape
^{\rm C}
       contained 37 different variables, many of which
С
       are superfluous to this study.
C
C.
       read(2,11)
      *flag, yr, day, hr, brn, idimf, idsw,
      *ptsi,ptsp,b,f,thb,phb,bxe,bye,bze,bym,
      *bzm, sigb, sigf, sigbx,
      *sigby, sigbz, t, n, v, phv, thv, sigt, sign, sigv,
      *sigphv,sigthv,kp,c9,r,dst
^{\circ}
       The averaging subroutines are called for the
C.
       various variables which have now been read in.
       These subroutines simply add the
C
       newly read variable to the previous sum of the
```

```
same variable and calculate a new average.
(,
      subroutine is called for each
٠.
      hourly reading until the last reading of the day
      is read in. At that time, the daily average value
      is computed and read into the file swdat as the
٠.
C
      average value for that variable for that day.
·
      call avgi(ptsi,nn(1),ss(1),aptsi)
      call avgi(ptsp,nn(2),ss(2),aptsp)
      call avgi(kp,nn(3),ss(3),akp)
      call avgi(c9, nn(4), ss(4), ac9)
      call avgi(r, nn(5), ss(5), ar)
      call avgi(dst,nn(6),ss(6),adst)
      call avgr(b,nn(7),ss(7),ab)
      call avgr(f,nn(8),ss(8),af)
      call avgr(thb,nn(9),ss(9),athb)
      call avgr(phb,nn(10),ss(10),aphb)
      call avgr(bxe,nn(11),ss(11),abxe)
      call avgr(bye,nn(12),ss(12),abye)
      call avgr(bze,nn(13),ss(13),abze)
      call avgr(bym,nn(14),ss(14),abym)
      call avgr(bzm, nn(15), ss(15), abzm)
      call avgr(t,nn(16),ss(16),at)
      call avgr(n,nn(17),ss(17),an)
      call avgr(v,nn(18),ss(18),av)
      call avgr(phv,nn(19),ss(19),aphv)
      call avgr(thv,nn(20),ss(20),athv)
C
      The 23rd hour of the day is the last reading on
C
      any given day. Thus at this time, the data must
C
C
      be finally averaged and then written into the file
^{\rm C}
      swdat. A total of 22 variables will be written
      into the new file.
C
      if(hr.eq.23) then
15
      format(i2,1x,i3,1x,2(f5.1,1x),f4.1,1x,
     *f3.1,1x,2(f6.1,1x),
     *9(f6.2,1x),f8.0,1x,f5.1,1x,3(f6.1,1x))
         write(3,15)
         yr,day,aptsi,aptsp,akp,ac9,
         ar, adst, ab, af, athb, aphb, abxe, abye,
         abze, abym, abzm,
         at, an, av, aphv, athv
      endif
      The following is the subroutine used to close the
٠.
      files when all the data from file3 has finally
\mathbf{C}
      been read and written. Note that this routine
      sets the termination variable dread equal to the
```

```
program termination value of 1.
      if(hr.eq.23.and.day.eq.101.and.yr.eq.85) then
         close(2)
         close(3)
         dread = 1
      end if
C
C
      The one data record which contained miscellaneous
      "junk" is identified with the routine below.
C
C
      This data is to be discarded.
                                      This statement was
C
      unique to the data provided.
С
      if(hr.eq.10.and.day.eq.338.and.yr.eq.82) then
          read(2,16) junk
16
          format(a180)
      endif
С
      A series of "if" statements to ascertain if the
c
      last original raw data record has been read.
C
      Again, the values used in these 3 statements are
C
      unique to the data provided.
C
C
      if(hr.eq.23.and.day.ne.101.and.yr.ne.85) goto 5
      if(hr.eq.23.and.day.eq.101.and.yr.ne.85) goto 5
      if(hr.eq.23.and.day.ne.101.and.yr.eq.85) goto 5
      goto 10
200
      end
C
      These are the actual subroutines used for
C
      averaging. Note that any variables which have a
С
С
      value of zero are not used in computing the final
      daily average. This is because tero was
С
      designated by the NSSDC as the value for a
С
С
      variable when its actual value was "missing" or
С
      indeterminate.
C
      subroutine avgi(x,nx,sx,ax)
      real sx,ax
      integer x,nx
      if (x.ne.0) then
        sx = x + sx
        ax = sx/nx
        nx = nx + 1
      endif
      end
```

```
subroutine avgr(x,nx,sx,ax)
real x,sx,ax
integer nx
if (x.ne.0) then
    sx = x + sx
    ax = sx/nx
    nx = nx + 1
endif
```



## Appendix B: FORTRAN Program Combining LANL and NSSDC Data

## program combol

This program was written to read the values in the files formed from the LANL data (energetic electron flux) and the NSSDC data (solar wind speed/IMF values) in order to form one single time synchronized data file for the purposes of performing transfer function analysis as described in Box and Jenkins. The name of the LANL file is "EEDAT3" while the name used for the NSSDC file was "swdat".

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C

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C

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C

In the following lists of variables, day, date, abzm (the average daily B<sub>z</sub> component expressed in Geocentric Solar Magnetospheric or GSM units), av (the average daily solar wind velocity in km/sec), and SEESSD, SEEI, SEEII, SEEIII, SEEIV (the various channels of energetic electron flux) are the variables of interest despite the fact that others were copied into the new time synchronized file. The other variables were either superfluous or declared for convenience in executing this program.

c c

C

c c

 $\mathbf{c}$ 

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 $\mathbf{C}$ 

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C

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C

C

The variables count and rep are counter variables used to control execution of this program. The variables doi and poi stand for "day of interest" and "port of interest". The 127th day of the file containing the solar wind and IMF data (file "swdat") represents May 8 which is the first day (matched to the file containing the energetic electron flux data, "EEDAT3") of sw/IMF values which should be copied. May 8, 1983 is the first day of data from file "EEDAT3". File "swdat" has data for days prior to this so all its data prior to this date is unecessary since no

cross correlations can be performed on it without data for similar days from file "EEDAT3". The parameter poi is a variable to name the port through which a file is copied (a FORTRAN particular when writing a file of data to a new file). Variables j1 and j2 are also counters.

count = 0 rep = 1 doi = 127 poi = 4 j1 = 0 j2 = 0

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c c

C.

Variable i is an array used to name the particular dates in file "EEDAT3" where missing values of variables of no interest to this study will have Thus, when the program determines to be read. that one of these dates is the date of the data being read, it will read the data according to a different format. In so doing, it will use the character variables junk1 and junk2 to read in the missing values of these useless variables. Without this caveat in the program, the creation of the new single time synchronized file could not proceed. Notice that there are 15 dates on which missing values of superfluous variables occur. The other 17 dates where missing values occurred were in August 1982. The reader will recall from Chapter III that the data up until May 8, 1983 from LANL was discarded partly for this reason (so many missing values in one month). Also, discarding data up to May 8, 1983 left exactly three years (1095 days) which could be analyzed. was felt to be more than enough data for analysis.

i(1) = 830523 i(2) = 830524 i(3) = 830526 i(4) = 840216 i(5) = 840217 i(6) = 840218 i(7) = 840620 i(8) = 840718 i(9) = 841015 i(10) = 841016 i(11) = 841031 i(12) = 841031

```
i(14) = 841102
      i(15) = 841103
\mathbf{c}
      The following statements open the relevant files.
C
      File "CCFLO" is the name given to the new single
C
      file containing combined, time synchronized
С
      sw/IMF and energetic electron flux data.
C
      reader will notice that later on in the program
C
      two other files ("CCFL1" and "CCFL2") were also
C
      created. These were initially thought to be necessary for the analysis but in fact were not.
C
C
      The name "CCFLO" is the author's acronym for
C
c
      "cross correlation file with 0 days lag".
      In other words, "CCFLO" represents a file of
C
      data from both "swdat" and "EEDAT3" which is
С
С
      time synchronized (ie., has zero days lag).
      open (2, file = 'swdat')
      open (3, file = 'EEDAT3')
      open (4, file = 'CCFLO')
      date = 830508
 5
      read (2,10) yr,day,aptsi,aptsp,akp,
                    ac9, ar, adst, ab, af, athb,
                    aphb, abxe, abye, abze, abym,
                    abzm, at, an, av, aphv, athv
С
       File "swdat" is read first since it has all
С
       the excess data at the start which must be
C
      discarded (ie., all data up until May 8, 1983).
C
C
 10
       format (i2, 1x, i3, 1x, 2(f5.1, 1x), f4.1, 1x, f3.1, 1x,
               2(f6.1,1x),9(f6.2,1x),
               f8.0, 1x, f5.1, 1x, 3(f6.1, 1x))
C
      We continue to read data from "swdat" but
C
      not copy it until May 8, 1983 rolls around.
C
      This is the first day of data from file "EEDAT3".
C
       When we finally reach this date, data from both
\mathbf{C}
       "swdat" and "EEDAT3" begins to be copied into
\mathbf{c}
       the file "CCFLO".
C
       if (yr.eq.83.and.day.eq.doi) then
            count = 1
       endif
       if (count.lt.1) goto 5
```

```
C
(,
      Continue to read data without copying until
C
      reaching the appropriate date.
C
C
      The following do loop takes care of reading
      in the appropriate values on those days when
C
C
      missing values for superfluous variables in
      file "EEDAT3" occur.
C
\mathbf{c}
      As mentioned earlier, without this the program
      could not read all the data. Notice that when
\mathbf{c}
\mathbf{C}
      the data is read this way, character variables
C
      junk1 and junk2 are assigned values for what
С
      would normally be the unused variables pts and
      GAMII (see statement 13 below).
C
C
      do 12 k=1,15
            if (date.eq.i(k)) then
                 j2 = k
                 k = 15
 10
                 read (3,11) grp,date,junk1,SEEIII,
                               junk2, SEEI,
                              SEEII, SEEIV, SEESSD
 11
                 format (i2, 2x, i6, 5x, a2, 1x,
                          f7.4,a7,3(f6.4,1x),f8.4)
            endif
 12
      continue
            if (j2.gt.j1) goto 16
C
С
      Statement 13 is the normal read statement used
      to read data from file "EEDAT3". Statement 10
С
      above is only used on the 15 particular dates
\mathbf{c}
C
      where missing values occur.
C
 13
      read (3,15) grp,date,pts,SEEIII,
                    GAMII, GAMIII, GAMIV,
                    SEEI, SEEII, SEEIV, SEESSD
 15
      format (i2, 2x, i6, 5x, i2, 4x, f6.4,
               3x, f7.4, 3x, f7.4, 2x, f8.4, 4x,
               f6.4,4x,f6.4,4x,f6.4,2x,f8.4
C
      Statement 16 begins the sequence where the data
С
      values from both "EEDAT3" and "swdat" are actually
C:
\mathbf{c}
      written into the new time synchronized file.
C
 16
      j1 = j2
      if (count.eq.1) then
            write (poi, 17) date, abxe, abye, abze,
```

```
abym, abzm, av, SEESSD,
                           SEEI, SEEII, SEEIII, SEEIV
            format (i6, 1x, 5(f6.2, 1x), f6.1, 1x,
 17
                    f8.4, 1x, 2(f6.4, 1x), f7.4, 1x, f6.1)
      endif
С
      The following list of statements are termination
С
      statements. The last day for which time synch-
С
      ronized data may be obtained is April 12, 1985.
C
      This is the last day of data occurring in file
C
      "swdat" and thus represents the last day for
С
      which we wish to write down values into the new
С
            April 12, 1985 corresponds to day 101 in
С
      the creation of file "CCFLO". It corresponds to
С
      day 100 and day 99 in the creation of files
c
      "CCFL1" and "CCFL2" respectively.
                                           These last two
С
      files were not used in the analysis. "EEDAT3"
C
      contains data for dates after April 12, 1985, but
C
      since "swdat" does not, we have copied as many
С
C
      values as we can for the purposes of transfer
Ċ
      function modeling.
C
      if (yr.eq.85.and.day.eq.101.and.rep.eq.1) then
           count = 0
            j1 = 0
           j2 = 0
           close (2)
           close (3)
           close (4)
            rep = 2
      endif
\mathbf{c}
      If we have not read the last day, then rep re-
С
      mains equal to 1 and we continue to read by re-
С
С
      turning to statement 5 to iterate the process.
      If we have read the last value, then rep is set
С
      equal to 2 (see the "if" statement above), and
С
      the new file is closed.
                                We then proceed on to
С
      create and fill data files "CCFL1" and "CCFL2"
С
      in the same manner as we did "CCFLO".
С
C
      if (rep.eq.1) goto 5
      if (count.eq.1) goto 20
      if (rep.eq.2) then
           open (2, file = 'swdat')
           open (3, file = 'EEDAT3')
           open (7, file = 'CCFL1')
           doi = 126
```

61.

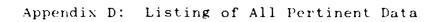
```
poi = 7
endif
20
     if (yr.eq.85.and.day.eq.100.and.rep.eq.2) then
          count = 0
          j1 = 0
          j2 = 0
          close (2)
          close (3)
          close (7)
          rep = 3
     endif
     if (rep.eq.2) goto 5
     if (count.eq.1) goto 25
     if (rep.eq.3) then
          open (2, file = 'swdat')
          open (3, file = 'EEDAT3')
          open (8, file = 'CCFL2')
          doi = 125
          poi = 8
     endif
25
     if (yr.eq.85.and.day.eq.99.and.rep.eq.3) then
          close (2)
          close (3)
          close (8)
          rep = 4
     end i f
     if (rep.eq.3) goto 5
     end
```

Appendix C: FORTRAN Program to Write Out All Pertinent Data

## Appendix C: FORTRAN Program to Write Out All Pertinent Data

```
PROGRAM DATSHO
        THIS PROGRAM WRITES OUT ALL THE DATA FOR EACH OF THE
        DAILY AVERAGE VALUES RECORDED FOR:
        SCLAR WIND, BZ, SEESSD SEEL, SEELL, SEELL, AND
        SEELV. THE BEGINNING DATE IS MAY 8, 1983. NOTICE
        THAT VALUES FOR THE SOLAR WIND ANT BE COMPONENT END
        ON 12 APR 85. THIS 706 DAYS REPRESENTS THE EXTENT
C
        OF THE CROSS-COBRELATABLE DATA BETWEEN THE SOLAR WIND/
        BZ CONSIDERED AS INPUTS AND THE INDIVIDUAL ELECTRON
        FLUX CHANNELS CONSIDERED AS CUTPUTS FOR TRANSFEE
        FUNCTION MODELING.
      INTEGER DATE, COUNT
      REAL ABZM, AV, SEESSD, SEEI, SEEII, SEEIII, SEEIV
      CHARACTER JUNE1, JUNE2, JUNE3
      COUNT = 1
      OPEN (2, FILE = 'CCFLO')
      OPEN (3, PILE = 'BBDAT3')
      READ (2,10) DATE, JUNEI, ABZM, AV, SEBSSD, SBEI, SEEII,
     *SEEIII, SEEIV
    FORMAT (16, A29, F6.2, 1x, F6.1, 1x, F8.4, 1x, 2(F6.4, 1x), F7.4, 1x. F6.4)
      IF (COUNT.EQ.1) THEN
         WRITE (6,11)
         FORMAT (3X,'DATE', 6X, 'BZ', 4X, 'SWVEL', 3X, 'SEESSE', 4X, 'SEEI',
 11
                4X, 'SEEII', 3X, 'SEEIII', 3X, 'SEEIV', /)
      ENDIF
      COUNT = COUNT + 1
      WRITE (6,12) DATE, AESM, AV, SEESSD, SEEL, SEELL, SEELL, SEEL
 12 FORMAT (2X, IE, 2X, F6.2, 2X, F6.1, 2X, F8.4, 2X, F6.4, 2X, F6.4, 2X,
                 F7.4,2X,F6.4)
      IF (COUNT.GB.45) THEN
         COUNT = 1
      BNDIF
      IF (IATE.LT.8EC412) GOTC 5
      CLOSE (2)
     REAT (3,15) JUNET, DATE
 15 FORMAT (A4.16)
```

```
IF (DATE LT.850412) GCT3 13
   BEA' (0,20) JUNE!, DATE, BUNES, SEBILL, JUNES, SEBI, SEBIL,
                 SEELY, SEESSI
25 FLEMAT (A4.16,5%,AC.4%,F6.4,A34,2(F6.4,4%),F6.4,2%,F8.4)
     IF (COUNT.EQ.1) THEN
        WEITE (5,22)
22
        FORMAT (3X, 'DATE', 6X, 'BZ', 4X, 'SWVEL', 3X, 'SEESSD', 4X, 'SEEI',
               4%, 'SEEII', 3%, 'SEEIII', 3%, 'SEEIV', /)
     ENDIF
     COUNT = COUNT + 1
     WRITE (6,25) DATE, SEESSD, SEEI, SEEII, SBEIII, SEEIV
25 FCRMAT (2X,16,18X,P8.4,2X,F6.4,2X,F6.4,2X,F7.4,2X,F6.4)
     IF (COUNT.GE.45) THEN
        COUNT = 1
     ENDIF
     IF (DATE.NE.860506) GOTO 17
    CLOSE (3)
     ENT
```

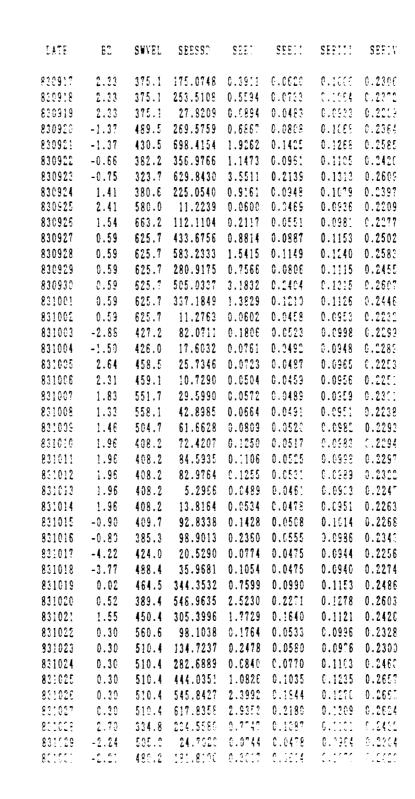


Appendix D: Listing of All Pertinent Data

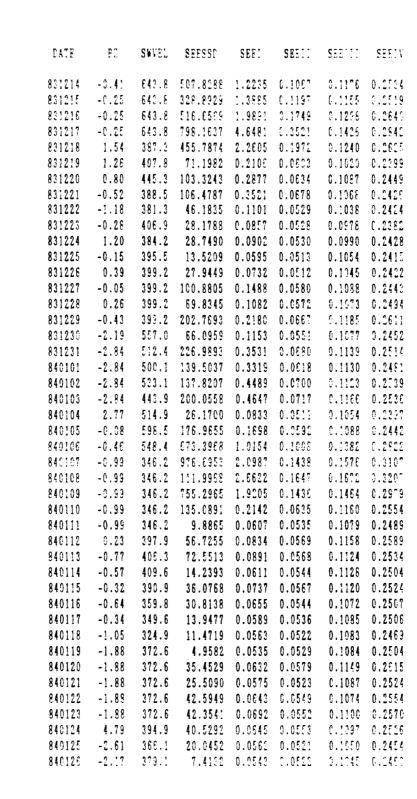
EATE	<b>5</b> .7	SWVE!	SEESSI	SEET	SEET	SEECLE	SEETT
800008	0.29	477.8	236.5678	8.4085	0.0841	0.0944	0.0000
830569	2.50	474.2	196.5375	0.4118	0.0800	0.0319	
830510	3.48	464.6	344.5045	1.0250	0.0908	0.0981	0.2151
830511	0.59	518.2	11.8411	0.0590	0.0397	0.0787	0.1880
830512	-1.02	£71.1	110.8708	0.3389	0.0557	0.0787	
830513	-1.02	671.1	282.7434	1.4830	0.1233	0.0968	0.1880
830514	-1.01	£71.1	429.1737	3.0454	0.2541	0.0953	0.1935
830515	-1.02	671.1	396.3473	2.7705	0.2319	0.0945	0.1939
830516	-1.02	671.1	374.1903	3.3052	0.3017	0.0981	0.1937
830517	-1.02	495.6	143.9721	1.0575	0.1227	0.0828	0.1802
830518	2.68	450.8	60.4168	0.5044	0.0692	0.0799	0.1794
	2.37	410.4	80.2263	0.5334	0.0722	0.0916	0.1897
	3.€3	389.4	187.6782	2.0617	0.2474	0.0968	
	5.82	557.5	3.655€	0.0399	0.0377		
830522	-3.05		31.3407	0.1020	0.0433	0.0800	
830523	-2.25	585.2	80.7859	0.4593	0.0611	0.0826	
830524	-1.33	614.2	184.3976	0.9290	0.1586	28.4798	
	5.95	648.0	184.4024	0.9290	0.1586	28.4781	0.2354
	5.95	648.0	55.8152	C.7328	0.1258	0.0938	
	5.95	648.0	184.4120	0.3291	0.1586	23.4747	0.2350
	5.95	648.0	70.4957	0.2087	0.0541	8.0895	0.1979
830529	5.95	648.0	152.8007	0.6294	0.0790	0.1948	5.2004
830530	0.86	391.3	119.905€	0.4777	0.0782	0.0905	0.2719
830531	0.79	420.6	20.3335	0.0631	0.0428	0.0864	0.1998
830601	-0.86	432.6	22.8509	0.0924	0.0469	0.0791	0.2048
830502	0.82	460.7	23.5162	0.0708	0.0431	0.0846	0.1984
830£03	0.31	502.0	34.9108	0.1309	0.0518	0.0915	0.2055
830504	0.51	438.9	45.3721	0.1657	0.0538	0.0830	0.2035
303068	0.89	440.9	55.9589	0.1887	0.0535	0.0889	0.2040
830606	2.72 2.72	517.3 517.3	55.6194	0.1645	0.0518	0.0365	0.1995 0.1999
830607 830608	2.72	517.3	29.4189 30.8373	0.0874 0.0895	0.0464	0.0865	0.1999
830609	2.72	517.3	15.6942	0.0588	0.0455	0.085€ 0.0831	C.1938
830610	2.72	517.3	4.6172	0.0386	0.0423	0.0833	0.1905
830611	2.72	357.0	15.6687	0.0478	0.0422	0.0829	0.1924
830612	2.72	365.2					
830613	-2.81	555.7	8.6462	0.0436	0.0391	0.0804	
830E14	0.84	463.5	87.8682	0.0696	0.0462	0.0573	
830615	-2.03	427.9	63.2878	0.0030	0.0402	0.0905	0.2068
830816	0.35	385.0	95.5263	0.1760	0.0507	0.0317	
830617	9.73	439.8	63.6880	0.1423	0.0461	0.0898	0.2019
839618	0.08	601.3	24.3370	0.0740	0.0444	1.1367	
830619	-1.73	592.4	73.4869	0.1230	0.0463	0.0338	0.2031
830620	-1.73	476.1	190.6555	0.4688	0.0594	0.0949	0.2000

DATE	<u>8</u> 2	SMAET	SEESSD	SEE!	SEEII	SEEIII	SEEIV
830621	-1.73	476.1	183.2236	0.4717	0.0664	0.0904	0.2108
830822	-1.73	47E.1	244.5261	0.5865	0.0644	0.0928	0.2065
830623	-1.73	476.1	176.3921	0.8379	0.070:	8880.0	0.2053
830624	-0.39	392.2	119.8399	0.5859	0.0734	0.0028	0.1064
830625	0.09	357.8	273.1826	1.7312	0.1023	0.1012	0.2139
830626	0.70	407.4	125.2135	0.3403	0.0805	0.0302	0.2068
830627	0.26	443.0	67.3353	0.2273	0.0506	0.0882	0.2027
830628	0.48	477.3	100.8528	0.3518	0.0625	0.0919	0.2053
830629	0.84	420.7	41.0516	0.0925	0.0453	0.0873	0.2034
830630	1.26	411.8	12.4443	0.0545	0.0438	0.0375	0.2037
839701	-0.32	393.5	48.8261	0.1317	0.0471	3063.0	0.2048
830702	-0.32	393.5	58.2352	0.1195	0.0472	0.0919	€.20€3
830703	-0.32	393.5	33.7165	0.0997	0.0458	0.0873	0.2065
830704	-0.32	393.5	24.0561	0.0522	0.0456	0.0878	[.210]
830705	-0.32	393.5	20.3643	0.0629	0.0485	0.0904	0.2129
830706	-0.32	363.9	8.3975	0.0534	0.0444	0.0891	0.2075
830707	-0.88	445.8	14.3057	0.0544	0.0455	0.0900	0.2120
830708	1.37	442.5	22.7939	0.0577	0.0445	0.0914	0.2103
830709	-0.39	455.3	19.5251	0.0672	0.0447	0.0916	0.2097
830710	0.78	465.0	51.3236	0.0929	0.0512	0.0947	0.2170
830711	0.60	399.7	57.6838	0.1071	0.0505	0.0969	0.2178
830712	2.07	401.5	43.1700	0.1060	0.0494	0.0938	0.2120
830713	-1.34	555.4	98.2269	0.0975	0.0515	0.0957	0.2176
830714	-1.34	555.4	125.5243	0.1024	0.0518	3860.0	0.2204
830715	-1.34	555.4	183.3351	0.1814	0.0548	0.0998	0.2223
830716	-1.34	555.4	3.7769	0.0461	0.0437	0.0954	0.2115
830717	-1.34	555.4	51.7207	0.0723	0.0479	0.0916	0.2141
830718	-1.34	555.4	56.9284	0.1151	0.0471	0.0916	0.2172
830719	-1.01	555.4	212.0317	0.3193	0.0575	0.0994	0.2205
830720	1.48	479.7	286.5022	0.6035 0.6114	0.0700	€.102€ 0.1013	0.2249
830721 830722	-1.03 -0.12	463.1 405.3	279.4203 124.2042	0.5023	0.0692 0.0627	0.1013	0.2181
830723	4.49	421.4	4.2854	0.0436	0.0627	0.0828	0.2161
830724	-3.85	443.1	21.2244	0.0676	0.0415	0.0883	0.1954
830725	-0.45	515.9	96.5466	0.1123	0.0489	0.0910	0.2123
830726	0.10	451.5	271.1768	0.3819	0.0628	0.1011	0.2123
830727	0.10	451.5	252.7928	0.3332	0.0628	0.099?	0.2246
830728	0.10	451.5	72.4116	0.1126	0.0504	0.0925	0.2153
830729	0.10	451.5	52.2086	0.0873	0.0476	0.0912	0.2150
830730	0.10	451.5	59.7341	0.1151	0.0483	0.0321	0.2161
830731	0.10	451.5	115.6035	0.1639	0.0527	0.0982	0.2248
830801	-0.86	272.2	192.3115	0.3423	0.0611	0.1011	0.224E 0.232E
800802	-0.66	427.2	5.8579	0.0460	0.0482	0.9936	0.2122
836810	2.68	564.0	4.6818	0.0478	0.0423	0,0898	1,0095

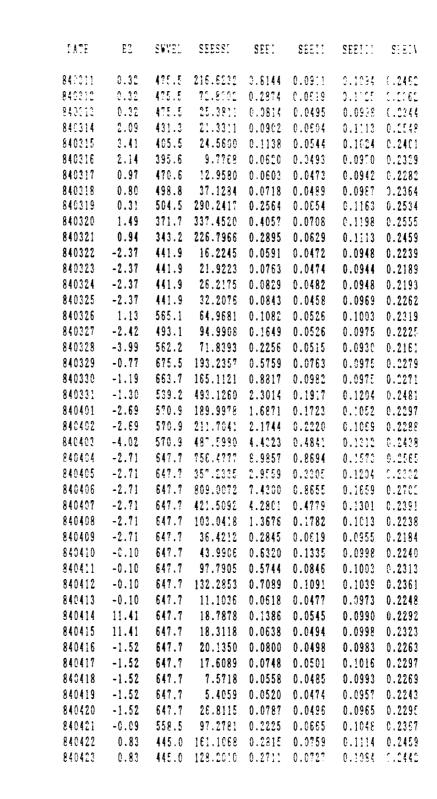
DATE	82	SWVED	SEESSD	SEEI	SEFII	SEFIII	SPEIV
830804	0.86	531.6	28.5173	0.0561	0.0472	0.0905	6.2145
890805	1.32	405.3	49.3039	0.0053	0.0522	0.0954	1,2305
830806	2.85	357.6	43.3183	0.0729	0.0475	0.0919	0.2081
830861	2.69	333.2	17.5633	0.0500	0.0434	((687	0.003:
830808	2.69	333.2	8.2698	0.0508	0.0442	0.0247	5.2012
830809	2.69	333.2	9.8932	0.0563	0.0442	0.0985	0.2035
830810	2.69	333.2	15.7431	0.0659	0.0429	3330.0	0.2051
830811	2.69	333.2	47.1030	0.1110	0.0475	0.0903	0.2092
830811	2.69	333.2	3.652€	0.0454	0.0426	0.0874	0.2083
830813	-0.63	542.5	22.4241	0.0648	0.0477	0.0920	0.2208
830814	1.05	509.7	132.0583	0.1283	0.0531	0.0996	0.2260
830815	0.24	491.2	183.3936	0.1504	0.0553	0.1000	0.2279
830816	0.06	427.3	250.4874	0.2351	0.0599	0.1034	0.2288
830817	1.34	381.2	38.4445	0.0813	0.0480	0.0914	0.2171
830818	1.78	351.4	212.9304	0.5943	0.0715	0.1060	0.2284
830819	3.17	355.6	135.3966	0.3975	0.0619	0.1001	0.2272
830820	-0.71	422.1	5.9777	0.0468	0.0455	0.0902	0.2145
830821	-0.71	422.1	8.6153	0.0476	0.0435	0.0901	0.2132
830822	-0.71	422.1	31.0290	0.0736	0.0478	0.0906	0.2234
830823	-0.71	422.1	42.9739	0.1049	0.0494	0.0925	0.2205
830824	-0.71	422.1	106.6274	0.2878	0.0587	0.0961	0.2222
830825	-0.71	422.1	180.5868	0.7948	0.0840	0.1035	€.2355
830826	-0.60	581.4	399.9948	1.9240	0.1498	0.1129	0.2362
830827	0.71	490.4	777.2976	5.5075	0.4246	0.1363	0.2643
830828	-1.46	415.6	310.0828	2.2563	0.2198	0.1111	0.2327
830829	0.36		53.2053	0.3127	0.0690	0.0918	0.2182
830830	0.62	635.3	6.8942	0.0510	0.0437	0.0863	0.2107
830831	-0.23	£39.9	166.2012	0.2020	0.0558	0.0986	0.2262
830901	0.70	582.8	196.4010	0.3766	0.0528	0.1024	0.2312
830902	0.70	515.0	282.1411	0.6869	0.0770	0.1073	0.2378
830903	0.70	515.0	250.7995	0.7646	0.0816	0.1066	0.2415
830904	0.70	515.0	292.0909	1.1802	0.1048	9.1090	0.2393
830905 830906	0.70 0.70	515.0 515.0	235.4409 86.8791	0.6026 0.2012	0.0821 0.0546	0.1067 0.0982	0.2365
830907	0.70	515.0	21.3002	0.2012	0.0346	0.0939	0.2292
830908	2.10			0.0523			0.2249
830909	2.49	362.0 528.0	5.9497 6.2765	0.0323	0.046E 0.0460	0.0951 0.0927	0.2251
830910	0.79	534.0	44.6622	0.0618	0.0514	0.0321	0.2275
830911	0.73	510.5	155.8239	0.1067	0.0568	0.1039	0.2339
830912	-0.15	465.8	96.7901	0.1413	0.0519	0.0995	0.2275
830913	-0.59	403.4	126.4273	0.2093	0.0565	0.1014	0.2318
830914	1.34	372.1	105.0611	0.2033	0.0535	0.0963	0.2267
830915	2.33	375.1	7.0178	0.0500	0.0444	0.0915	0.220
830316	2.33	375.1	13.6484	0.0567	0.0466	0.0914	0.2030 0.0010
	2100	0.011	10.0101		010766		



DATE	ВZ	SWVEL	SEESSD	SEEI	SEE	SEE!!!	SEETV
831021	-0.67	411.3	199.6836	0.5826	0.0732	0.1101	0.0468
801101	-1.75	370.3	127.0907	0.3452	0.0021	0.1054	0.2435
831102	2.82	611.9	13.2441	0.0587	0.0481	0.0987	1.2355
831103	1.39	€05.9	53.7797	0.0732	0.0506	0.1035	1.2169
831104	1.45	532.4	77.5977	0.1185	0.0534	0.1043	0.2425
831105	1.21	532.4	116.8316	0.1934	0.0807	0.1118	1.2432
83110€	2.25	532.4	158.953?	0.286€	0.0639	0.1058	0.2435
831107	-1.23	532.4	50.6835	0.0994	0.0522	0.0960	0.2281
831108	-2.97	532.4	23.9970	0.0580	0.0468	0.0919	0.2238
831109	-1.62	589.1	82.2664	0.1600	0.0511	0.0948	1.2228
831110	-1.10	574.8	270.4604	0.5886	0.0749	0.1065	0.2364
831111	4.25	590.3	107.2401	0.4689	0.0648	0.0936	3.2190
831112	-3.83		54.8620	0.1577	0.0512	0.0912	0.2145
831113	-2.74	415.5	36.4082	0.1052	0.0486	0.0346	9.2211
831114	-1.10	478.8	110.6407	0.3547	0.0607	0.1001	0.2293
831115	1.58	535.9	98.6521	0.4022	0.0634	0.1022	0.2345
831116	0.14	622.8	166.9456	0.3532	0.0681	0.1023	0.2357
831117	0.53	735.0	307.0937	1.0932	0.1060	0.1096	0.2411
831118	0.34	735.0	594.0546	1.9180	0.1552	0.1216	0.2574
831119	1.05	735.0	609.1188	2.4430	0.1901	0.1246	0.2605
831120	0.17	735.0	827.3979	3.6066	0.2750	0.1376	0.2705
831121	0.36	735.0	767.1244	3.7039	0.2723	0.1415	0.2786
831122	0.78	469.3	836.2339	4.6821	0.3899	0.1487	0.2854
831123	1.57	385.9	419.1855	4.1379	1.0926	0.1979	3.3287
831124	-0.72	370.5	314.9636	1.0887	0.1309	0.1154	0.2504
831125	-1.07	444.4	77.0917	0.1306	0.0549	0.0973	0.2369
831126	-1.26	440.0	123.2007	0.2365	0.0575	0.1044	0.2372
831127	0.72	405.1	101.5761	0.1952	0.0590	0.1033	0.2441
831128	2.22	448.2	42.4152	0.1278	0.0548	0.1013	0.2414
831129	1.47	618.2	19.2049	0.0739	0.0498	0.1041	0.2426
831130	0.68	618.2	115.4216	0.1257	0.0559	0.1058	0.2458
831201 831202	0.90 1.27	618.2	278.6878 320.9668		0.0581 0.0738	0.1172	0.2518
831203	1.65	618.2 618.2	361.7669	0.4582 0.6689	0.0842	0.1159 0.1222	0.2580
831204	1.14	618.2	327.3109	0.7183	0.0887	0.1222	0.2562
831205	-0.86		60.4529				0.2385
831206	-1.65	595.6	54.1124	0.1102	0.0530	0.1042	0.2368
831207	-1.11	610.7	331.9057	0.5205	0.0748	0.1187	0.2550
831208	0.38	523.4	30.6408	3.3498	0.2054	0.1524	0.2997
831209	3.56	441.4	891.6741	4.5244	0.2699	0.1324	0.2331
831210	0.13	422.8	268.4093	1.3646	0.1215	0.1118	€.245€
831211	-0.49	460.8	8.7918	0.0557	0.0472	0.0350	0.2200
831212	0.62	562.2	36.4771	0.0889	0.0487	0.0961	0.2271
831213	0.10	643.8	175.2748	0.2484	0.0590	0.1039	0.2341

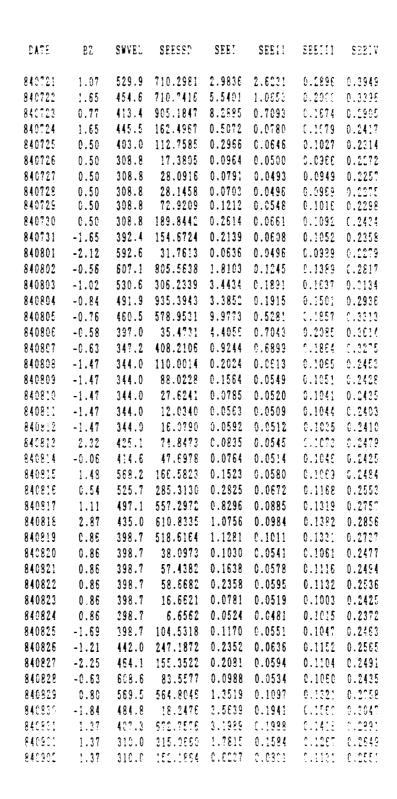


DATE	ВZ	SWVEL	SEESSI	SEEI	SEEII	SPELLI	SEEIV
840127	-0.94	385.5	5.8702	0.0548	0.0537	0.1084	0.2515
840128	0.99	483.5	6.8249	0.0523	0.0513	1.1065	0.2513
840129	1.08	491.8	10.4035	0.0544	0.0505	0.1013	6.2427
840130	-0.36	671.2	81.5637	0.1238	0.0540	0.1040	0.2425
840131	0.58	668.6	263.6094	0.4902	0.0732	0.1131	0.2562
840201	(.73	658.8	354.4764	0.9795	0.0980	3.1229	0.2633
840202	0.73	65.8	403.9552	1.1240	0.1031	0.1278	0.2699
840203	0.73	658.8	31,8193	0.0993	0.1031	0.1022	0.2079
840203	0.73	658.8	10.0612	0.0533	0.0483	0.0988	0.2339
840205	0.73	658.8	32.8101	0.0907	0.0515	0.1017	0.2370
84020E	-0.10	398.2	87.3635	0.1818	0.0595	0.107	0.2483
840207	0.48	394.5	41.1644	0.0976	0.0551	0.1065	0.2475
840208	-0.51	391.7	142.3114	0.2823	0.0658	0.1000	0.2608
840209	-0.47	428.0	66.9102	0.1093	0.0558	0.1064	0.2508
840210	-2.81	416.6	90.7658	0.1551	0.0570	0.1091	0.2487
840211	-2.53	430.8	65.6584	0.0387	0.0554	0.1083	0.2450
840212	-0.18	395.1	138.9470	0.1535	0.0611	0.1169	0.2603
840213	-6.37	392.1	34.2986	0.0873	0.0498	0.1030	0.2361
840214	-6.37	392.1	82.1503	0.2044	0.0556	0.1033	0.2413
840215	-6.37	392.1	48.4983	0.1556	0.0546	0.1048	0.2427
840216	-6.37	392.1	88.5640	0.3162	0.0935	0.1585	0.3596
840217	-6.37	457.7	230.8326	1.0715	0.143?	11.7252	1.1250
840218	-0.57		230.8374	1.0716	0.1437	11.7234	1.1251
840219	0.38	412.6	230.8421	1.0716	0.1437	11.7217	1.1252
840220	3.82		13.3758	0.0722	0.0509	0.1048	0.2333
840221	-1.09	463.5	29.2731	0.0741	0.0534	0.1043	0.2424
840222	0.26	428.5	45.3264	0.0777	0.0548	0.1076	0.2474
840223	0.50	437.7	29.7743	0.0641	0.0515	0.1048	0.2414
840224	2.53	560.5	34.3807	0.0673	0.0509	0.1006	0.2386
840225	1.16	434.8	54.2461	0.0938	0.0546	0.1088	0.2457
840226	0.39	386.5	27.3344	0.0653	0.0505	0.1041	0.2414
840227	0.39	388.5	13.2964	0.0595	0.0503	0.0987	0.2316
840228	0.39	388.5	36.5885	0.0720	0.0510	0.1008	0.2336
840223	0.39	386.5	40.4755	0.0699	0.0518	0.1013	0.2343
840301	-1.58	544.2	22.3254	0.0585	0.0495	0.0972	0.2271
840302	-0.95	681.7	98.1633	0.2382	0.0553	0.0970	0.2241
840303	-1.20	592.9	554.5409	1.6796	0.1266	0.1213	0.2523
840304	0.32	467.1	£48.0546	3.4300	0.2262	0.1323	0.2670
840305	2.10	406.5	481.4972	3.5431	0.7833	0.1847	0.3220
840305	-1.33	551.6	213.7771	1.3684	0.1145	0.1082	0.2423
840307	-1.21	538.0		0.4261	0.0658	0.1025	0.2368
840309	-1.26	545.4		1.4383	0.1141	0.1172	0.2482
849309	0.32	475.5	146.9998	1.1759	0.1265	0.1058	0.2355
840316	0.32	475.5	190.8641	0.9703	0.0371	0.1056	0.2373

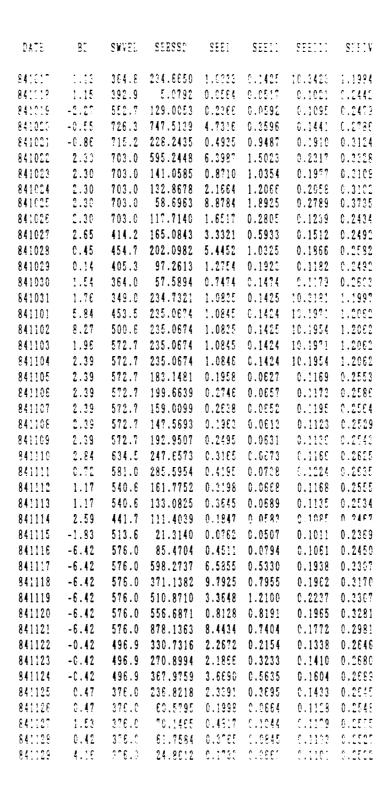


DATE	<b>B</b> 2	SWVEL	SEESSE	SEEI	SEETI	SEEIII	SEEIV
840424	1.00	445.0	26.5909	0.0672	0.0521	0.0994	6.2341
840425	-3.34	445.0	13.2240	0.1513	0.0928	0.1243	0.2403
84042E	-3.72	445.0	148.6642	2.0717	0.7248	0.4:32	0.3453
840427	-1.33	533.8	298.6903	1.3784	0.3613	0.2004	0.2595
840428	-1.33	533.8	774.4427	3.4992	0.2778	0.1556	0.2478
840429	-1.33	533.8	39.3608	5.2853	0.3139	0.1515	0.2639
840430	-1.33	533.8	98.7328	6.0495	0.3098	0.1540	0.2783
840501	-1.33	533.8	45E.3110	2.0570	0.1350	0.1195	0.2005
840502	-1.33	533.8	327.2597	2.0885	0.1883	0.1196	0.2387
840503	0.69	474.8	366.1742	3.4422	0.3385	0.1247	0.2014
840504	-1.84	458.2	161.1268	0.6010	0.0734	0.1050	0.2295
840505	-3.20	400.4	128.3955	0.2701	0.0573	0.1034	0.2284
840506	0.81	362.4	38.7974	0.1561	0.0527	0.1023	0.2275
840507	2.60	335.3	91.5988	0.4279	0.0589	0.1075	0.2311
840508	4.86	371.1	103.5002	0.4574	0.0861	0.1026	0.2279
840509	0.15	385.0	36.2406	0.1385	0.0518	0.0983	0.2227
840510	2.12 6.25	434.0	11.8988 9.2150	0.0541	0.0459	0.0983 0.0923	0.2200
840511 840512	6.25	464.8 418.9	5.5503	0.0312	0.0437	0.0923	0.2066 0.2115
840512	6.25	418.9	9.0273	0.0477	0.0437	0.0924	0.2116
840514	6.25	418.9	10.6299	0.0533	0.0446	0.0994	0.2140
840515	6.25	418.9	15.5008	0.0539	0.0464	0.0982	0.2191
840516	6.25	418.9	26.6945	0.0533	0.0479	0.1007	0.2255
840517	3.53	538.5	11.1762	0.0555	0.0444	0.0918	0.2049
840518	-0.42	471.8	10.1852	0.0535	0.0460	0.0929	0.2100
840519	-0.44	454.1	39.4121	0.1112	0.0473	0.0931	0.2133
840520	-0.78	483.8	80.2955	0.1149	0.0518	0.1002	0.2205
840521	-2.23	541.3	31.2427	0.1534	0.0518	0.0994	0.2218
840522	-1.63	590.1	362.9964	1.4700	0.1128	0.1132	0.2375
840523	0.52	636.2	621.6406	4.1070	0.2812	0.1274	0.2510
840524	2.19	635.6	475.0843	3.3244	0.2731	0.1157	0.2409
840525	2.19	635.6	158.5559	0.7110	0.0756	0.1003	0.2267
840526	2.19	635.€	210.1347	1.0432	0.1021	0.1061	0.2192
840527	2.19	635.6	464.4539	2.7071	0.1960	0.1248	0.250€
840528	-1.63	480.0	323.2195	1.5239	0.1251	0.1132	0.2395
840529	-0.86	438.1	164.1316	0.5700	0.0704	0.1039	0.2248
840530	3.81	418.5	64.5959	0.3225	0.0595	0.0976	0.2219
840531	2.66	418.5	19.4648	0.1268	0.0514	0.1016	3222.0
840601	1.70	334.5	63.7952	0.4468	0.0671	0.1637	0.2338
840602	-0.76	390.5	36.3599	0.1131	0.0503	0.1007	0.2250
840603	-0.44	471.5	37.2352	0.1522	0.05:9	0.0389	0.2212 9.2183
840604	-0.70	566.8	26.8641	0.0680	0.0489	0.1395	0.0000
840605	-1.23	588.7 FOC 7	92.2160	0.1435	0.0526	0.1631	0.2219
840606	-1.23	588.7	156.2100	0.2337	0.0554	0.1659	0.2311

DATE	FZ	SWVEL	SEESSI	SEEI	SESII	SEE	SELIV
940607	-1.20	588.7	129.9914	0.1922	0.0556	0.1058	0.2311
840609	-1.23	588.7	192.9457	0.4599	0.0657	0.1069	
840609	-1.23	588.7	43.7022	0.1053	0.0509	0.1004	
840£10	-0.59	656.0	216.7139	0.2214	0.0596	0.1194	0.2428
840811	0.40	587.6	600.4174	0.5800	0.0319	0.1294	0.2643
840612	0.08	512.3	659.7035	0.9450	0.0975	0.1234	0.2670
840613	0.32	365.5	579.0411	1.0772	0.0959	0.1363	0.2695
840614	C.07	367.1	377.3713	0.6673	0.0897	0.1239	0.2631
840615	1.78	490.5	74.4494	0.1341	0.0541	0.1045	0.2381
840616	-0.22	646.8	74.1187	0.1388	0.0549	0.1028	0.2289
840617	1.39	530.1	143.3897	0.2233	0.0574	0.1051	0.2337
840618	1.39	509.8	84.0062	0.1980	0.0546	0.1013	0.2265
840619	1.39	509.8	126.4588	0.1946	0.0549	0.1035	0.2278
840620	1.39	509.8	113.4334	0.2436	0.0580	0.0950	0.2375
840621	1.39	509.8	232.7680	1.0775	0.1431	11.0268	1.1621
840622	1.39	509.8	542.5692	3.5022	0.2316	0.1322	0.2631
840623	1.39	509.8	565.5732	7.3005	0.5081	0.1324	0.2608
840624	1.12	453.7	65.1381	0.2589	0.0582	0.1062	0.2361
840625	0.32	408.2	66.5555	0.2664	0.0615	0.1077	0.2399
840626	0.60	408.0	109.9318	0.3920	0.0696	0.1095	0.2436
840627	-0.69	379.6	78.7414	0.1637	0.0565	0.1068	0.2425
840628	-4.56	376.3	14.0224	0.0669	0.0477	0.1021	0.2341
840629	2.12	442.6	7.1928	0.0523	0.0492	0.1002	0.2285
840530	-2.33	504.3	79.3223	0.0817	0.0530	0.1052	0.2341
840701	-2.33	504.3	112.6738	0.1244	0.0546	0.1062	0.2375
840702	-2.33	504.3	86.6237	0.0975	0.0503	0.1031	0.2358
840703	-2.33	504.3	138.8681	0.1451	0.0546	0.1064	0.2367
840704	-2.33	504.3	112.4972	0.1416	0.0545	0.1038	0.2331
840705	-2.33	572.5	124.6674	0.1597	0.0578	0.1092	0.2392
840706	-0.47	541.3	154.7901	0.2073	0.0592	0.1122	0.2411
840707	0.16	514.0	204.5325	0.2585	0.0635	0.1161	0.2515
840708	-0.20	451.6	241.4356	0.2353	0.0630	0.1188	0.2544
840709	-0.26	434.6	163.1024	0.1798	0.0589	0.1121	0.2522
840710	-1.20	427.7	118.4210	0.1393	0.0558	0.1060	0.2360
840711	-2.80	382.9	80.8242	0.0983	0.0549	0.1030	0.2350
840712			19.1828		0.0488		
840713	1.41	419.0	4.2970	0.0497	0.0465	0.0946	0.223?
840714	1.41	419.0	135.9740	0.3468	0.0610	0.1053	0.2354
840715	1.41	419.0	478.2426	1.2106	0.1034	0.1230	0.2593
840716	1.41	419.0	559.9037	2.6790	0.1679	0.1245	0.2627
840717	1.41	419.0	939.4742	5.7021	0.3803	0.1469	0.2824
840718	1.41	419.0	928.7066	4.3264	0.3085	0.1412	0.2912
840719	1.0€	419.0	233.2374	1.0789	0.1430	10.8574	1.1711
840720	-0.20	554.5	69€.7854	6.5673	2.0402	0.2855	0.3957



DATE	5.5	SWVEL	SEESSI	SEE!	55511	SEEIII	SE: IA
840903	1.37	310.0	51.9340	0.1871	0.0001	0.1688	0.2499
840304	1.37	310.0	10.8360	0.0771	0.0590	0.1032	2.24.6
840905	1.37	310.0	37.2016	0.1810	0.6554	0.1020	0.2080
840906	-0.56	310.0	275.2681	0.994€	0.1399	0.1165	1.0561
840907	1.62	410.0	342.3694	1.3632	0.1719	0.1225	0.2584
840908	4.94	401.8	427.1530	3.1328	0.0852	0.1241	0.2647
840909	2.41	478.1	274.3305	0.6577	0.0911	0.1183	0.2617
840910	0.53	702.1	143.595?	0.4275	0.0734	0.1112	0.2551
840911	0.49	693.9	453.5367	1.0391	0.1041	0.1307	0.2729
840912	0.52	€35.0	487.8800	1.0487	0.1051	0.1315	0.2786
840913	2.93	614.9	994.2604	2.8085	0.1793	0.1604	0.3087
840914	2.93	614.9	61.1532	4.3661	0.2810	0.1599	0.3111
840915	2.93	614.9	986.8851	4.2315	0.2843	0.1611	0.3111
840916	2.93	614.9	445.4474	1.8700	0.1576	0.1318	0.2732
840917	2.93	614.9	448.0508	2.3083	0.2086	0.1372	0.2777
840918	2.93	614.9	707.6599	5.7569	0.4912	0.1581	0.2931
840919	-0.87	522.3	58.5614	0.4342	0.0887	0.1087	0.2458
840920	-0.68	448.5	225.4315	0.2329	0.0639	0.1120	0.2547
840921	-0.46	460.2	383.4935	0.7006	0.0802	0.1221	6.6021
840922	-0.23	462.9	317.5739	0.6457	0.0765	0.1205	0.2643
840923	-2.38	643.9	76.3074	0.2481	0.0590	0.1025	0.2426
840924	-0.49	751.4	941.1225	6.0505	0.4564	0.1578	0.2911
840925	-0.72	680.0	603.8993	5.0378	1.2639	0.2135	0.3335
840926	-2.21	685.0	46.9794	4.3674	1.4728	0.2041	0.3010
840927	-2.21	685.0	781.5458	9.2272	1.0231	0.1796	0.2892
840928	-2.21	685.C	27.8259	9.1128	0.9936	0.1929	0.0033
840929	-2.21	685.0	398.5352	4.9359	C.7368	0.1819	6.2727
840930	-2.21	685.0	286.2952	2.0504	0.2092	0.1282	0.2663
841001	-2.21	685.0	414.4194	5.7008	0.7369	0.1663	0.2708
841002	-1.20	336.0	238.4428	1.6975	0.2316	0.1270	0.2817
841003	1.79	372.7	16.5669	0.0729	0.0524	0.1051	0.2441
841004	1.99	368.9	8.4053	0.0641	0.0516	0.1060	0.2440
841005	2.25	356.0	11.4484	0.0648	0.0518	0.1045	0.2409
841005	1.20	403.3	7.7531	0.0549	0.0476	0.0953	0.2269
841007	3.00	556.3	14.7853	0.0640	0.0492	0.1018	0.2416
841008	0.97	698.3	183.6185	0.2446	0.0836	0.1126	0.2511
841009	0.97	710.5	291.1329	0.4274	0.0726	0.1191	0.2627
841010 841011	0.97 0.97	710.5 710.5	710.3834	2.2427	0.1570 0.2009	0.1428 0.1496	0.2919
841012	0.97	710.5	124.0873	5.3760	0.2809	0.1996	0.2063
841013	0.97	710.5	98.1200	5.8271	0.3607	0.1692	0.3134
841014	-1.80	536.0	367.0674	7.1343	0.5042	0.1652	0.3340
841015	-0.65	471.3	985.9666	6.6835	0.4981	0.1988	0.3115
841016	-1.40	458.8	234.6632	1.0833	0.1425	10.3441	1.1984
5 T = 1 1 =	1175	700.0	B 6 7 6 0 6 0 E	. 1 4 5 0 0		1010771	141307



\*27.5

DATE	ВZ	SWVEL	SE5581	SEEI	SEEII	SERILL	SEELY
841130	4.16	378.1	13.8331	0.0945	6.0517	0.1080	0.2415
841201	4.18	30€.0	191.2490	0.1187	0.0012		
841000	4.18	378.0	£2.££10	30805	0.0000		
841003	4.16	37E.C	440.1147	0.5454	0.0767	1.1518	
841204	4.16	378.0	326.5432	0.8984	0.0807		
	0.10	707.1	361.6308	0.9742	0.0955	0.1208	
	0.08	644.0	886.9927	2.6387	0.1839		
841207	-0.19	635.2	741.2852	2.8107	0.1919	0.1423	(.299)
841208	-0.57	635.2	964.6821	4.0155	0.2598	6.1719	0.3166
841209	-0.57	635.2	861.5584	3.7093	0.2414	0.1693	3.0208
841210	1.27	368.5	557.2236	2.4619	0.2023	0.1432	0.2856
841211	-1.44	562.5	191.9280	0.5519	0.0853	0.1182	C.2582
841212	0.83	521.6	62.2337	0.1079	0.0574	0.1119	0.2502
841213	0.83	521.6	104.0586	0.1146	0.0575	0.1116	0.2501
841214	0.83	521.6	293.3018	0.2818	0.0701	0.1229	0.2648
841215	0.83	521.6	479.8995	0.6725	0.0867	0.1349	0.2744
841216	0.83	521.6	312.5101	0.4379	0.0744	0.1225	0.2660
841217	-0.52	521.6	643.2272	1.3805	0.0144	0.1223	0.2810
841218	-0.49	598.7	282.4524	3.7688	0.2362	0.1771	0.3255
841219	-0.58	469.4	555.9530	4.7805	0.2953	0.1771	0.3596
841219	-1.51	401.3	577.4872	5.9143	0.2333	0.2944	0.3602
841221	1.14	398.0	627.3108	2.8222	0.2118		0.2864
841222	0.03		58.0347	0.1204	0.0578	0.1451 0.1103	0.2517
841223	-1.10	445.0	10.7553		0.0530	6.103E	0.2451
	0.64	445.0		0.05£3			0.2549
841224	3.95		61.7885	0.0928	0.0581	0.1129	0.2548
841225	3.95	320.6	29.7314	0.0677	0.0539	0.1095	
841226	3.95	320.6	16.2548	0.0605	0.0518	0.1062	0.2461
841227	3.95	320.6 320.6	32.0501	0.0690	0.0532	0.1082	
841228	3.95		245.9883	0.2571	0.0673	0.1213	0.2592
841229		320.6	399.0954	0.5095	0.0765	0.1281	0.2768
841230	3.95 3.95	619.3	765.9008	1.5992	0.1202	0.1453	0.2917
841231		661.6	768.5547	1.9921	0.1382	0.1454	6.2924
850101	3.95	724.1	467.5444	1.8427	0.1449	0.1362	0.2714
850102	3.95	682.9	303.7992	1.0191	0.1085	0.1278	0.2739
850103	3.95	682.9	783.8506	1.9083	0.1637	0.1540	0.2944
850104	3.95	453.0	924.7980		0.2206	1.1590	0.3155
850105			548.8491	1.9472	0.1653		0.2955
850106			171.7523	0.2654	0.0723		
850107		424.6	180.0259	0.3767	0.0741	0.1167	0.2660
850108	3.95	424.6	77.1919	0.2306	8330.0	0.1139	0.2608
850100 850110	3.95		121.5398	0.1812	0.0597	0.1129	0.2591
850110	3.95	424.6	855.5011	1.5252	0.1250	0.1550	0.3016
850111	0.92		808.1394	1.7332	0.1329		0.3092
850112	1.32	629.0	974.5895	2.5763	C.1675	0.1654	0.3179

DATE	25	SWVEL	SEESSI	SEEI	SEEII	SEEIII	SEELV
850110	0.57	574.4	973.1761	3.6460	0.2218	0.1651	0.3218
850014	-0.60	574.4	999.0408	4.5013	0.2741	0.1899	0.5191
850115	0.99	446.5	518.3029	2.2621	0.1500	0.1347	0.2925
850116	U.36	484.7	200.0687	0.7731	0.0890	0.1215	0.2681
850117	-0.10	412.1	184.7342	0.7658	0.0875	0.1216	0.2638
850118	0.57	370.2	45.6031	0.2027	0.0649	0.1115	0.2580
850119	-0.54	342.5	8€.2874	0.4023	0.6773	0.1152	0.2569
850120	-0.54	342.5	63.3030	0.2495	0.0693	0.1132	0.2554
850121	-0.54	342.5	78.2433	0.2468	0.0684	0.1133	0.2597
850122	-0.54	342.5	15.6787	0.0687	0.0564	0.1079	0.2525
850123	-0.54	600.0	47.7974	0.0860	0.0550	0.1103	0.2530
850124	-0.10	469.3	175.6067	0.1863	0.0670	0.1236	0.2687
850125	6.75	479.0	191.2083	0.2278	0.0627	0.1164	0.2548
850126	1.65	409.8	139.899?	0.2255	0.0631	0.1118	0.2466
850127	-2.52	399.7	48.2095	0.0696	0.0557	0.1100	0.2466
850128	-4.72	490.0	46.5021	0.0948	0.0533	0.1060	0.2464
850129	-2.42	481.1	200.3278	0.6910	0.0839	0.1222	0.2673
850130	0.81	469.3	512.2911	1.9704	0.1592	0.1358	0.2862
850131	-0.55	417.3	463.0419	1.9725	0.1475	0.1380	0.2841
850201	-0.55	417.3	337.8270	1.2648	0.1093	0.1390	0.2781
850202	-0.55	417.3	30€.3029	1.4144	0.1337	0.1303	0.2761
850203	-0.55	417.3	608.1980	3.7692	0.2897	0.1442	0.2940
850204	-0.55	417.3	530.8087	4.2241	0.3293	0.1453	0.3054
850205	-0.55	529.8	65.0849	0.3064	0.0708	0.1117	0.2500
860206	2.31	529.8	38.6712	0.0871	0.0547	0.1988	0.2510
850207	1.62	675.5	295.1098	0.3172	0.0721	6.1061	0.2717
850208	0.25	647.4	502.4382	0.7589	0.0895	0.1356	0.2904
850203	0.71	593.1	382.1718	1.2338	0.1049	0.1331	0.2820
850210	2.65	665.0	581.5863	2.4078	0.1683	0.1452	0.2951
850211	0.93	627.4	633.6258	3.0455	0.2253	0.1491	0.3010
850212	0.71	559.2	810.8228	3.3673	0.2741	0.1646	0.3266
850213	1.02	510.3	831.9723	4.7919	0.3904	0.1655	0.3308
850214	1.02	510.3	347.0404	1.392?	0.1463	0.1310	0.2839
850215	1.02	510.3	242.3249	0.8622	0.1077	0.1286	0.277€
850216	1.02	510.3	311.7280	1.8292	0.1920	0.1319	0.2848
850217			172.0926				
850218	1.55	438.1	14.1167		0.0952	0.1177	0.2672
850219	0.88	391.3	37.8612	0.7063	0.1067	0.1145	
850220	-2.24	331.0	24.5211		0.0589	0.1115	
850221 850222	-2.57	376.0 364.9	24.EE71	0.0982	0.0594	0.1165	
850222		349.3			0.0579 0.0576	0.1128 3.1139	
850224			26.8173 6.8921		0.0510		
850225	-1.78 -1.59	412.2		0.0596	0.0572	0.1122	0.2082
01077	-1.07	764.6	43.:035	V.VI.C	0.00.2	·	4 . 5

PATE	<b>E</b> 2	SWVEL	SEESSI	SEE!	SEEII	SEEIII	SEETV
850226	-2.40	375.5	48.7117	0.0781	0.0581	0.1147	0.2854
850227	-2.42	375.5	36.0171	0.0749	0.0571	0.1112	0.2657
850228	-2.42	375.5	110.0338	0.1716	0.0898	0.1182	0.1881
850301	-0.40	375.5	222.4541	0.4592	0.0753	0.1271	0.2020
850302	1.34	598.0	326.9134	0.8003	0.0897	0.1234	0.2548
850103	1.65	804.0	234.9399	0.5975	0.0801	0.1244	0,0770
850304	2.43	462.4	231.4010	0.7187	0.0953	0.1279	0.2784
850305	1.22	658.7	173.5347	0.4921	0.0772	0.:212	5,2000
850306	1.97	701.6	378.2263	0.8363	0.0995	0.1294	1171.0
850307	1.68	760.6	428.0356	0.9491	0.1033	0.104:	1,1860
850308	1.45	694.6	768.8057	3.5889	0.2599	0.1577	0.3114
850309	0.93	504.6	323.2904	9.2491	0.7062	0.2017	0.3538
850310	5.60	485.0	348.4363	3.2427	1.2118	0.2188	0.2421
850311	5.60	465.0	72.1522	0.2236	0.082?	0.1161	0.2612
850312	5.60	465.0	75.1244	0.4148	0.0762	0.1135	0.2512
850313	5.60	465.0	18.6538	0.1004	0.0566	0.1125	0.2616
850314	-1.03	383.0	24.4017	0.1264	0.0588	0.1138	0.2620
850315	1.23	466.4	8.9149	0.0607	0.0537	0.1107	0.2593
850316	0.93	454.7	12.5553	0.0675	0.0557	0.1113	0.2601
850317	1.12	404.0	29.5768	0.0857	0.0590	0.1145	0.2681
850318	-0.82	393.2	11.7745	0.0638	0.0562	0.1131	0.2625
850319	-1.67	375.8	12.8501	0.0602	0.0546	0.1094	0.2828
850320	1 19	363.4	17.2438	0.0517	0.0575	0.1114	0.2646
850321	v.11	331.8	16.1537	0.0631	0.0563	0.1110	0.0858
850322	0.53	333.2	7.3158	0.0595	0.0572	0.1156	0.2678
850323	-0.55	318.2	6.0041	0.0575	0.0548	0.1137	0.2630
850324	-0.55	318.2	4.3811	0.0562	0.0548	0.1112	0.2628
850325	-0.55	318.2	3.9757	0.0579	0.0535	0.1120	0.2850
850326	-0.55	318.2	4.3546	0.0551	0.0538	0.1108	0.2584
850327	-2.63	327.4	3.6204	0.0558	0.0546	0.1106	0.2564
850228	2.38	378.9	3.0909	0.0588	0.0556	0.1095	0.2819
850329	2.33	405.4	3.7493	0.0580	0.0571	0.1136	0.2001
850330	2.54	435.1	19.6132	0.0615	0.0580	0.1143	0.2094
850331	3.39	492.5	8.3199	0.0580	0.0557	0.1119	0.2014
850401	1.08	582.0	3.4146	0.0514	0.0546	0.1099	0.2666
850400	1.08	692.5	66.3900	0.0735	0.0598	0.1157	0.2680
			360.9408		0.0810	0.1331	0.2981
850404	1.08		739.3507		0.1116	0.1659	0.3044
850405		552.8	772.7992	1.5073	0.1235	0.1840	0.3350
850406		553.8	799.3716	2.0656	0.1490	0.1702	0.3411
850407		\$53.8	93.7434	0.1501	0.0844	0.1232	0.2794
850403			57.0671	0.1386	0.0621	0.1196	0.2110
850419			28.9248		0.0572	0.1164	0.2700
		433.0	50.5119	0.0778	0.0593	0.1190	1371.)

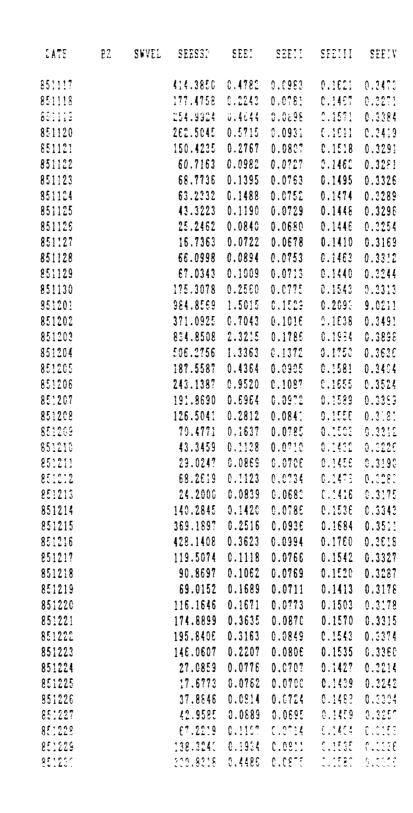
DATE	<b>P</b> .2	SWVEL	SEESS	SEEI	SEFII	SEE	SEEEV
850411	1.08	477.5	42.9496	0.0807	0.0615	0.1208	0.0742
850412	1.08	469.1	62.6717	0.1176	0.0651	3.1146	0.1901
850413			82.6152	0.1729	0.0669	0.1288	0.2825
850414			118.1039	0.1493	0.0648	0.1228	0.0825
850415			81.0242	0.1495	0.0656	0.1252	0.2839
850416			57.5142	0.0980	0.0605	0.1234	0.2783
850417			35.9637	0.0755	0.0599	0.1215	0.2802
850418			55.1824	0.0891	0.0629	0.1259	0.2854
850419			10.2484	0.0597	0.0567	0.1158	0.2733
850420			5.6966	0.0547	0.0553	0.1111	0.2644
850421			44.5212	0.2207	0.0653	0.1168	0.2684
850422			149.5794	0.7759	0.1159	0.1271	0.2856
850423			334.6781	1.3693	0.1443	0.1370	0.2948
850424			333.9201	1.6086	0.1572	0.1477	0.3023
850425			387.6622	1.9593	0.1697	0.1435	0.2999
850426			83.5531	0.2692	0.0652	0.1108	
850427			172.4058	0.5695	0.0735	0.1118	
850428			193.6247	1.4132	0.1825	0.1124	0.2398
850429			152.947?	1.0986	0.1226	0.1145	
859430			77.0905	0.8475	0.1357	0.1140	
850501			27.6922	0.1054	0.0564	0.117€	0.2472
850502			120.9975	0.2376	0.0660	0.1129	0.2904
850503			227.0148	0.4756	0.0815	0.1190	
850504			165.7690	0.2751	0.0688	6.1186	0.2999
850505			52.2372	0.2382	0.0727	0.1145	
850506			48.7417	0.1147	0.0584	9.1153	
85050°			46.6091	9.1129	0.0609	0.1167	
850508			43.3762	0.1452	0.0611	0.1170	
850509			19.3634	0.0690	0.0568	0.1121	
850510			10.8347	0.0663	0.0576	0.1187	0.2708
850511			15.3406	0.0651	0.0575	0.1170	
850512			8.3345	0.0605	0.0554	0.1146	0.2693
850513			14.8606	0.0667	0.0583	0.1169	0.2730
850514			4.4752	0.0592	0.0565	0.1179	0.2745
850515			16.2300	0.0604	0.0559	0.1158	0.2705
850516			59.1014	0.0758	0.0604	0.1191	0.2773
850517			20.0846	0.0644	0.0570	0.1193	0.2737
850518			9.7911	0.0591	0.0581	0.1165	0.2713
850519			3.9184 2.8543	0.0587	0.0587	0.1177	0.2764
850520 850521			3.3998	0.0608 0.0597	0.0598 0.0581	0.1209	0.2922
850522			4.4134	0.0597	1.0599	0.1138	0.2810 0.2812
850523			5.1982	0.0608	0.0601	0.1218 0.1213	0.2838
850524			4.4861	0.0607	0.0614	0.1213	0.2620
Culul4			4.4351	0.0001	0.0514	U `	4.60

DATE	<b>E</b> ?	SWVEU	SEESSE	SEEI	SEELI	SEETTI	SEEIV
850525			4.2090	0.0599	0.0593	0.1194	0.2783
850526			4.3002	0.0588	0.0578	0.1171	
850527			7.2409	0.0635	0.0606	0.1227	
850528			9.7236		0.0594	0.1221	
850529			5.5385	0.0626	0.0598		
850530			6.4009	0.0659	0.0638		
850531			4.1216		0.0601	0.1180	
850601			F 2152	0.0585	0.0589	0.1183	
250602			E.3153 54.4520	0.0727	0.0629	0.1274	
850503			90.2557	0.0834	0.0875		
850604			55.6228	0.0731	0.0621		
850605			29.8984	0.0649	0.0604	0.1224	
850606			0.709/	V V L U 3	0.0582		
850607			118.2839	0.1077	0.0653		
803038			95.0725	0.1778		0.1253	
850609			147.2381	0.3874	0.0743		
850510			141.1535	0.1923	0.0677		
850611			209.7101		0.0760		
850612			437.5824	0.9058	0.0987		
850613			456.6597	1.4198	0.1217		
850614			116.3140	0.2269	0.0652	0.1397	0.3027
850615			169.3583	0.3825	0.0784	0.1314	0.2948
850616			42.7678		0.0660	0.1276	0.2850
850617			33.0527	0.1203	0.0658	0.1259	0.2905
850618			19.4851		0.0606		
850619			26.1073		0.0638		
850620			17.1960	0.0742	0.0611	0.1220	
850621			11.1866		0.0604		
850622			13.7543	0.0681	0.0619	0.1245	
850623			8.7151	0.0685	0.0592	0.1216	
850E24			7.3900	0.0701	0.0629	0.1242	
850625			13.3375	0.0862	0.0590	0.1179	
850E26			23.3168	0.0818	0.0611	0.1214	
850627			121.4709		0.0686		
850628			66.1696	0.1020	0.0633	0.1233	0.2869
850629			233.9606		0.0765	€.1355	0.0007
850630			354.0147	0.6815	0.0911	0.1459	0.3070
850701			78.9412	0.1458	0.0674	0.1285	0.2921
850702			98.6462	0.2035	0.0718	0.1299	0.3022
850703			150.1100	0.4464	0.0842	0.1356	0.3010
950704			21.3993	0.0761	0.0574	0.1175	0.2800
850705			72.7435	0.1044	0.0005	0.1214	
860708			236.6770	0.3104	0.0742	0.1225	
850707			428.3557	1.0878	0.1050	1.1476	0.1134

DATE	EI	SWVEL	SEESSI	SEET	SEEII	SEEDLY	SEEIV
850708			818.6736	2.2598	0.1958	3.1748	9.3050
850709			E16.2460	2.0351	0.1598	1,1684	(.:445
850710			727.6400	3.5100	0.2474	0.1783	6.3444
850711			157.4796	0.810?	0.1107	0.1295	0.295:
850712			85.9040			0.1162	
850713			213.5669			0.1238	
850714			87.9351			0.1215	
850715			708.4595	3.0264	0.1954	0.1680	7.1198
850716			85.5039	8.7911	0.4683	0.1933	0.3690
850717			211.9707	0.5944	0.0883	1.1360	0.29€€
850718			175.8679	0.4416	0.0787	0.1309	0.2943
850719			287.9275	0.5853	0.0860	0.1392	0.4751
850720			301.1668	0.7409	0.0974	0.1435	0.3134
850721			118.8429	0.2359	0.0729	0.1292	0.2985
850722			110.8537	0.3979	0.0778	0.1297	0.2984
850723			24.1796	0.1040	0.0619	0.1201	0.2851
850724			31.1936		0.0630		0.2880
850725			45.5564	0.0892	0.0628		
850725			112.3856	0.1762	0.0694		
850727			184.9409	0.3289	0.0753		
850728			274.1722		0.0871		
850729			326.9473		0.0893		0.3263
850730			132.9214		0.0702		9.2918
850731			32.5911			1.1189	
850801			152.7215	0.2811	0.0708	0.1270	0.2878
850802			223.3760	0.3751	0.0785	11114	1.2909
850803			357.1899			0.1473	
850804			546.1469			0.1650	
850805			764.2014	2.6444		0.1774	
85080€			751.5192			2.1841	
850807			346.5358			0.1462	
850808			78.2761			0.1266	
850809			60.3367			0.1267	
850810			25.4305	0.0807	0.0603		
850811			32.8004	0.0969		0.1290	
850812			22.4294		0.0602	0.1210	
850813			4.0082	0.0577	0.0560	0.1212	4.8584
850814			543.6551	1.1686	0.1142	0.1529	0.3192
850815			825.8046	2.3924	0.1693	0.1752	0.3386
850816			350.6109	0.9839	0.1044	0.1490	0.3141
850817			388.3256	0.8564	0.0942	0.1468	0.0163
850818			152.5599	0.7224	0.0271	0.1327	0.3014
850319			138.2944	0.3298	0.0740	0.1324	0.3007
850820			113.9978	0.2730	0.0709	9.1308	0.2903

DATE	B 2	SWVEL	SEESSD	SEEI	SEEII	SEEIII	SEELV
859821			207.5400	0.4647	0.0854	0.1398	0.3081
850822			203.1777	0.4735	0.0801	0.1369	0.3028
850823			609.6134	1.9841	0.1615	0.1579	0.3354
850824			529.7751	1.3347	0.1232	0.1628	0.3341
850825			37.0343	0.1411	0.0844	0.1288	0.2327
850826			50.5888	0.1286	0.0888	0.1255	0.2338
850827			82.6192	0.1238	0.0648	0.1269	0.2943
850828			164.3920	0.1724	0.0717	0.1340	0.2982
850829			240.4725	0.3366	0.0796	0.1372	0.2991
850830			441.1464	0.8818	0.1084	0.1574	0.2163
850831			44.5041		0.0620	0.1231	0.2871
850901			114.7793	0.2667	0.0751	0.1375	€.30€5
850902			260.3091		0.1040	0.1478	0.3203
850903			334.1283	0.9737	0.1078	0.1515	0.3291
£50904			353.7156	1.2182	0.1259	0.1544	0.3004
850905			197.4968			0.1450	
850906			40.7174	0.1097	0.0686	0.1306	0.2995
850907			7.6278	0.0641		0.1257	0.2978
850908			14.1384			0.1262	0.2985
850909			23.5899		0.0813		0.29გა
850910			69.0698	0.0920		0.1314	0.3032
850911			174.8101		0.0735		0.3123
850912			160.5024		0.0743		0.3160
850913			164.8553			0.1407	
850914			22.6247	0.0727	0.0621	0.1238	0.3012
850915			3.6805	0.0584	0.0854	0.1310	0.3069
850916			95.0488	0.1138	8330.0	0.1321	
850917			455.0642	0.4789	0.0960	0.1630	
850918			709.2010	1.3322	0.1319		
850919			87.8466	0.2132	0.9677	0.1301	
850920			286.1552	0.3903	0.0825	0.1401	
850921			320.0055	0.8112	0.0947		
850922			740.0520	2.1408	0.1604	0.1705 0.1607	0.3557
850923 850924			517.9929 447.1563		0.1467	0.1607	0.6581
850925			276.6996		0.1521	0.1090	
850926			579.2537				
850927			762.2559				
850928			615.5533			0.1767	
850929			688.3818			0.1784	
850930			850.1644			0.2012	
851001						0.1898	
851002			242.9973				0.2151
851003			42,5328		0.000		0.2024
0.1000			7414040	0.1666		⊍ <b>1.4</b> ,	

DATE	27.	SWVEL	SEESSO	SEEI	SFEII	SEELII	SEEIV
851004			5.6245	0.0814	0.0613	0.1284	0.3002
851005			5.6245 45.9420	0.1016		6.1261	
851006			832.6738	3.3734		0.1768	
851007			585.6230	8.4722		0.2388	
851008			195.1833	5.7377		0.2923	
851009					2.1193		0.4842
851010			736.6350	6.6455	1.4784		0.4437
851011			226.2078 736.6350 100.5311 59.4001 175.1964 222.7244 125.1821 181.8630	0.2744		0.1362	0.3078
851012			59.4001	0.2251	0.0709	0.1349	0.3998
851013			175.1964	0.4707	0.0839	0.1422	0.3117
851014			222.7244	0.4752	0.0883	0.1452	0.3181
851015			125.1821	0.3364	0.0779	0.1358	0.3137
851016			181.8630 287.5163	0.3465	0.0773	0.1426	0.3162
851017			287.5163	0.4143	0.088?	0.1468	0.3267
851018			136.7514	0.2723	0.0735	0.1388	0.3151
851019			287.5163 136.7514 513.9333 828.7290	0.7831	0.1103	0.1671	0.3464
851020					0.1675	0.1896	
851021			469.3183		0.1350	0.1641	
851022			98.6132	0.1643	0.0720		
851023			470.1825	0.6705	0.1053	0.1714	
851024			459.5151		0.1113	0.1609	
851025			344.8286		0.1012	0.1574	
85102E			188.9425	0.4433	0.0919	0.1531	
851027			311.3256		0.1224		
851028			127.5457		0.0797		
851029			45.9771	0.1070			
851030			4.0212	0.0654			
851031			5.5642	0.0681	0.0656	0.1355	
851101			9.1739	0.0664	0.0646		
851102			21.5432	0.0703	0.0657	0.1042	
851100 051104			191.7515	0.1927	0.0775	0.1468	
851104 851105			982.0961	1.2127	0.1377	0.2016	
851106			699.5004		0.1254	0.1808	0.3656
851107			243.7546	0.6416 0.5282	0.0900		0.2335
851108			264.7381		0.0893	0.1557	
851108			241.3829	0.6654	0.0976	0.1506	0.2298
851109 851110			96.6271	0.1712	0.0743 0.0725	0.1429	0.3209
851111			180.1557	0.1290	0.0809	0.1441 0.1523	0.3316
851112			172.5565	0.2176	0.0815	0.1523	0.3336
851112			44.2855	0.2409	0.0661	0.1301	0.3336
851114			30.0637	0.0002	0.0654	0.1379	0.3095
851115			179.3899	0.1595	0.0034	0.1474	0.3251
851116			538.8126	0.5149	0.1030	0.14.4	0.3574
301110			00000	0.0173	0000	711100	0.00:4



DATE	EI	SWVEL	SEESS?	SEEI	SEEII	SEE	SPEIV
851231			861.0352	1.9677	0.1878	0.0004	((()))
860101			758.3288	2.2590	0.1772	1,1571	
201038			24.277	3.7009	0.2728	1.215	[,4,]7
860103			4:0.2384	5.4683	0.0519	2,202.0	0.4470
860104			393.1720	8.5279	0.5792	0.2425	9.4304
860105			216.8599	7.9458	0.8829	(. <b>2</b> 000	1.0107
86010E			€08.1573	5.0139	0.4716	0.1860	0.3685
860167			66.8608 157.2783		0.0712		
860108			157.2783	0.1500	0.0822		
860109			352.3887	0.3002	0.0967		
860110			143.8157	0.1615	0.0802		
860111			223.6698	0.2820	0.0893		
860112			165.3522	0.2117	0.0856		
860113			118.5711	0.1855	0.0803		
860114			76.3317		0.0797		
860115 860116			43.8961		0.0744	0.1454	
860117			17.2557		0.0711	0.1474	
860118			10.0397		0.0677		
860119			10.3231		0.0704	0.1409	
860120			5.6771	0.0637	0.0654		
860121			21.3216	0.0761	0.0646	0.1366	
860122			70.7506	0.0840	8830.0		
860123			210.3508		0.0797		
860124			690.6239		0.1137		
860125			455.5797	1.1098	0.1119		0.2401
860126			319.0968	0.5946	0.1010		
860127			407.2978	0.8201	0.1050		
860128			999.0785	2.7040	0.2091	0.2034	
860129			738.9543	7.7203	0.4804	0.2527	0.4811
860130			573.9995	6.9938	0.4452	0.2504	0.4459
860121			729.5627	8.8391	0.5869	0.2646	0.4595
860201			574.2798		0.7201		
850202			659.0290			0.1939	
860203			374.7841			0.1698	
860204			264.3569	1.2609	0.2053	9.1654	0.3431
860205			187.9661	0.9605	0.1715	0.1575	0.3400
860206			111.0884	0.7420	0.1916	0.2114	0.3745
860207			25.4355	0.2218	0.1375	0.1839	0.3264
860208 860209			91.6344	0.4604	0.0918	0.1235	0.2683
			36.4915	0.1708	0.0682	0.1095	0.2492
860210 860 <b>2</b> 11			44.5956 110.2774	0.2979	0.0398 0.0395	0.1244	0.2722
860212			109.3230	0.4667	0.0893	0.1275 0.1210	0.2754 0.275
500514			103.3631	0.1103	0.0511	0.1410	V . L : L

DATE	BC	SWVEL	SEESSC	SELI	SEELL	SEE:::	SEELY
860210			831.9343	2.1004	0.1527	0.1883	0.3456
860214			707.6456			0.2190	
800215			420.0558		0.1834		
860216			381.7810		0.3360	0.1470	
860217				1.0423	0.1639		
860218			48.3821	0.1256	0.0602	0.1182	
860219			75.1698			0.1234	
860220			75.1698 41.1743 66.2223 280.3640 567.4124 64.6458	0.1122	0.0618	0.1210	
860221			66.2223	0.1181	0.0649		
860222			280.3640	0.2710	0.0824	0.1400	
860223			567.4124	1.2964	0.1219	0.1599	
860224			64.6458 461.5604	2.9596	0.2311	0.2041	0.3882
860225			461.5604	4.8787	0.3282	0.230€	0.4213
860226			433.4131	0.0303	V.41U3	0.2130	0.4034
860227			974.3854	4.7552	0.3577	0.2056	0.3819
860228			660.6248	3.6639	0.3172	0.1784	0.3443
860301			297.8015	1.6004	0.1750	0.1551	0.3175
860302			279.8955	1.3175	0.1525	0.1512	0.3238
860303			352.3002	5.0322	0.7346	0.1916	0.3290
860304			167.6465	0.6603	0.1088	0.1410	0.21E0
860305			199.3438	1.4770	0.2029		0.3166
850305			63.3244			0.1315	0.2008
702038			99.3302			0.1306	0.3021
860308			213.4366	0.8040	0.0900	0.1088	0.3079
860369			197.4989	0.6674	0.0979	0.1005	€.289€
860010			544.0297	4.6521	0.3093	0.1900	r 900r
860311			711.6551	5.4215	0.3620	0.1739	0.3391
860312			210.2005	1.0368	0.1230	0.1368	0.3002
860313			33.7488	0.0901	0.0586	0.1231	0.2849
860314			41.8412	0.0908	0.0624	0.1261	0.2956
860315			34.8689	0.0949	0.0623	0.1253	0.2932
860316			45.6402	0.085€	0.0633	0.1291	0.2967
860317			51.7143	0.1065	0.0639	0.1300	0.3029
860318			48.0447	0.0904	0.0643	0.1281	0.2991
860319			71.7350	0.1022	0.0663	0.1323	0.3090
860320			88.6467		0.0724	0.1363	0.3180
860321			28.7168	0.0780	0.0608	0.1285	0.3009
860322			104.7850	0.1304	0.0685	0.1307	0.3047
860323			165.9279	0.1644	0.0736	0.1377	0.3119
860324			137.0872	0.2069	0.0710	0.1347	0.3033
860325			235.2775	0.4311	0.0834	0.1464	0.3134
860326			506.6770	1.1258	0.1219	0.1099	0.3487
880327			385.5760	0.8884	0.1054	0.1559	0.2312
850328			E61.0150	1.4488	0.1418	0.1670	0.0573

DATE	BC	SWVEL	SEESSD	SEEI	SEEII	SEE	SEELV
860009			407.2204	1.0387	0.1222	0.1585	0.3349
850330			238.8583	0.8584	0.1071	3.1516	0.0294
860531			180.2263	0.6585	0.0998	(.1400	0.3210
860401			119.3902	0.3756	0.0842	0.1411	0.0171
860402			105.7562	0.4482	0.0953	0.1420	0.3174
860403			34.0527	0.1260	0.0693	(.1044	9.0109
860404			42.8514	0.1826	0.0738	0.1421	0.3218
860405			58.4321	0.2725	0.0788	(.:279	33985
860406			19.933?	0.0866	0.0701	0.1383	0.3177
860407			15.2751	0.0784	0.0655	0.1049	
860408			12.9174	0.0844	0.0689	0.1255	0.3165
860409			5.8188	0.0696	0.0668		
860410			10.8256	0.0669	0.0655	0.1314	0.3098
860411			23.1654	0.0753	0.0687	0.1385	0.3163
860412			25.8039	0.0691	0.0639	0.1331	0.3106
860413			13.2699	0.0589	0.0677	0.1367	0.3137
860414			18.5467	0.0694	0.0658	0.1341	0.3179
860415			5.5622	0.0653	0.0542	0.1294	0.3145
860416			8.2890	0.0679	0.0666	0.1354	0.3176
86041?			8.7771	0.0693	0.0678	0.1352	0.3161
860418			10.0942	0.0673	0.0671	0.1323	0.3180
860419			7.3609	0.0644	0.0624	0.1332	0.3140
860420			11.0364	0.0691	3330.0	0.1350	0.3154
860421			11.5644	0.0702	0.0671	0.1379	0.3208
860422			7.0411	0.0646	0.0649	0.1342	0.3160
860423			9.7012	0.0683	0.0688	0.1350	0.0208
860424			31.2800	0.0769	0.0679	0.1360	0.3184
860425			28.3760	0.1228	0.0753	0.1405	0.3271
860426			26.2116	0.1012	0.0733	0.1409	0.3263
860427			23.6217	0.0948	0.0712	0.1424	0.3263
860428			4.9680	0.0685	0.0860	0.1364	0.3225
860429			15.7443	0.0721	0.0683	0.1421	0.3288
860430			26.9121	0.0769	0.0712	0.1419	0.3327
860501			24.5988	0.0761	0.0701	0.1421	0.2215
860502			8.8463	0.0666	0.0646	0.1334	
860503			66.2500	0.2120	0.0734	0.1423	
860504			113.85€7	0.3063	0.0799	0.1474	0.3295
860505			106.8020	0.1375	0.07??		
860506			6.6490	0.0727	3330.0	0.1333	0.3115

## Appendix E: BMDP Example Problems

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BMDP	Example Problem #2, Possible Good Model for a Series Under Study	144
BMDP	Example Problem #3, Typical Transfer Function Model for Two Interacting Series .	150

## Appendix E: BMDP Example Problems

BMIF Example Froblem #1

Computer: SSC

Data set used: SHEEDATYE (7 May 85 - 6 May 86)

Model fit attempted: AR1,27

Poor model

BMDPOT - BOX-JENEINS TIME SERIES ANALYSIS

Mon Aug 4 16:56:30 1980

/PROBLEM TITLE IS 'GROSYNCH ENERGETIC ELECTRON TIME SERIES'.

/INPUT VARIABLES ARE 10.

FORMAT IS FREE.
FILE IS SHEEDATYR.

RECL = 95.

/VARIABLE NAMES ARB DATE, PTS, SEBIII, GAMI, GAMIII, GAMIV, SBBI,

SEEII, SEEIV, SEESSD.

/BND

ALL CASES ARE COMPLETE.

THE BLOCKING IS ACROSS ALL VARIABLES.

ACF VARIABLE IS SEEL.
MAXLAG IS 60./

FIRST CASE NUMBER TO BE USED = 365
LAST CASE NUMBER TO BE USED = 365
NC. OF OES. AFTER DIPPPRENCING = 365
MEAN OF THE (DIPPERENCES) SERIES = C.9742
STANDARD ERROR OF THE MEAN = 0.1212
T-VALUE OF MEAN (AGAINST ZBEC) = 8.0465

## AUTO MERFOATIONS

.- :2 19. 30. 40. 40. 80. 80. 80. 60. 60. 81. 80. 30. SI.F. 13- 24 22. 88.- 38.- 39.- 30.- 38.- 88.- 18.- 18.- 18. 28. 48.-ST.E. 25- 36 .08 .13 .15 .12 .08 .05 0.0 -.02 -.03 -.05 -.06 -.06 ST.E. 20. E0. 37- 48 -.05 -.03 -.03 -.04 -.04 -.07 -.08 -.08 -.08 -.08 -.08 -.16 ST.E. 49-60 -.11 -.09 -.06 -.03 .01 .04 .05 .03 0.5 -.02 -.05 -.07 ST.E.

#### PLOT OF AUTOCORRELATIONS

-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 C.8 1.0

CORR. +----+----+ LAG 1 0.758 + IXX+XXXXXXXXXXXXXXXXXXXXX 2 0.428 + IXXX+XXXXXXX 0.200 + IXXX+X 4 0.050 - IX + 5 -0.003 3 0.000 0.017 8 0.049 - <u>ΤΣ</u> - • ĝ 0.067 [XX + 10 0.075 + IXX + 0.060 11 + IXX + 16 0.057 IX . 13 0.042 + IX + 14 0.020 15 0.000 + IX + 16 0.009 17 -0.013 18 -0.035 19 -0.002 + 🟋 🖁 🚶 20 -0.061 + XXI 21 -0.057 + 11 12 -0.053 + XI 23 -0.027 + XI 24 0.018 25 0.085 + IXX + 28 0.134 · [XXX-

```
0.148
                            · [XXXX
    0.119
                            · [XXX:
   0.079
                            · !!! ·
    0.000
                            · !X ·
+ X:
33 -0.032
                            + 11
34 -0.054
                            + YI
                            + %:
35 -0.053
36 -0.061
                            + 111
37 -0.043
                            + 11
38 -0.028
                            + 11
39 -0.033
                            + 11
40 -0.03E
                            + 🟋 [
41 -0.044
                            + XI
42 -0.067
                            + 177
43 -0.091
                            + XXI
44 -0.083
                            + 111
45 -0.079
                            + 111
46 -0.078
                            + 111
47 -0.073
                            + XXI
48 -0.097
                            + %%[
49 -0.107
                            + % % % 1
50 -0.090
                            + 111
51 -0.054
                            + 111
52 -0.025
                            + 71
53 0.012
54 0.037
                            + 17 +
55
   0.053
                               IX
56 0.035
57 -0.001
58 -0.023
                            + 11
59 -0.046
                            • XI
60 -0.068
                            + XXI
```

# PACE VARIABLE IS SEEL. MAXLAG IS 60./

FIRST CASE NUMBER TO BE USES	:	2
LAST CASE NUMBER TO BE USED	:	365
NO. OF OBS. AFTER DIFFERENCING	:	388
MEAN OF THE (DIFFERENCE!) SERIES	=	0.9742
STANDARD BREOR OF THE MEAN	=	0.:2:0
T-VALUE OF MEAN (AGAINST ZERO)	:	5.0411

## PARTIAL AUTOCIERGUATIONS

1- 12 - 178 - 38 177 - 163 116 - 162 116 117 117 117 117 117 117 117 33 31, 10, 30, 10, 30, 30, 30, 30, 30, 30, 30, 30, ST.E. 31. 30. 1.0 20.-101 10.-301-101 30.-101 30. 301-10 13- 24 39. 39. 30. 30. 30. 30. 30. 30. 30. 30. 30. ST.E. 25 - 08 \$20.+ 20.- 80.- 30.- 30. \$0.- 10. \$0. \$0.- 20. \$0. .CE .GE .GE .OE .OE .GE .GE .GE .GE .GE .GE ST.E. 27- 48 .03 -.04 -.03 0.0 -.03 -.04 .01 0.0 -.02 0.0 -.03 -.05 ST.E. 49- 60 20.- 40.- 10. 10.- 20.- 10.- 04 - 01 - 03 - 01 - 04 - 02 

## PLOT OF PARTIAL AUTOCORRELATIONS

-1.0 -0.8 -0.5 -0.4 -0.2 0.0 0.2 0.4 0.£ 0.8 1.0 Ţ 1 0.758 + [XX+XXXXXXXXXXXXXXXXXXX 2 -0.348 3 0.075 • IXX• 4777 + 4 -0.095 5 0.098 · [XX-€ -0.016 - T -7 0.023 + [] + 8 0.04? + ! 1 -+ 1 -9 -0.010 10 0.039 + [] + 11 -0.041 + 77 + 12 0.076 + []]+ 10 -0.081 + X X I + + 17 + 14 0.021 15 0.033 + 17 + 16 -0.063 + X X I + 17 0.012 + I + 18 -0.057 + 11 + 19 -0.009 20 0.014 + 1 + 21 -0.040 4 77 + 22 0.004 + : + · [] · 23 0.031 24 0.054 + 11 + 25 0.080 + [XX-

```
26 0.018
2.
    0.019
28 -0.020
   0.025
30 0.010
31 -0.073
                              + X X I +
30
   0.046
                              + [] +
33 -0.059
                              + 11 +
34 -0.025
                              + 11 -
35 -0.018
                              + I +
3€ -0.034
                             + 11 +
37 0.033
                              + 1% +
38 -0.035
                              + 11 +
39 -0.032
                              + 11 +
40 0.004
41 -0.030
                              + XI +
42 -0.043
40 0.008
                              + I +
44 0.001
45 -0.019
46 -0.004
47 -0.030
                             + 11 +
48 -0.054
                             4 | | | +
49
   0.021
                             + IX +
50 -0.020
                             + XI +
51 0.010
                             + [ +
52 0.022
                              + IX +
53 -0.013
                              + I +
54 0.045
55 -0.009
56 -0.034
57 -0.008
58 0.010
                             + I +
59 -0.039
                             + 11 +
60 -0.025
                             + 11 +
```

ARIMA VARIABLE IS SEEI.
AROR = '(1),(27)'./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS OUTPUT VARIABLE = SEEI INFUT VARIABLE = NOISE

ESTIMATION RESIDUAL = RESEEL.

FIRST CASE NUMBER TO BE USED = 1 LAST CASE NUMBER TO BE USED = 005

ESTEMATION BY CONTITIONAL LEAST SQUARES METELL

RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN 0.1000e-04

SUMMARY OF THE MODEL

OUTFUT VARIABLE -- SEE!
INPUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFERENCES

SEEI BANDOM 1- 365

 PARAMETER VARIABLE
 TYPE
 FACTOR
 ORDER ESTIMATE
 ST. ERR.
 T-RATIO

 1
 SBBI
 AR
 1
 1
 0.785€
 0.0339
 23.15

 2
 SEEI
 AR
 2
 27
 0.1040
 0.0545
 1.91

RESIDUAL SUM OF SQUARES = 836.986938
DEGREES OF PREEDOM = 335
RESIDUAL MEAN SQUARE = 2.493468

ESTIMATION BY BACKCASTING METHOD

RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN 0.10008-04

SUMMARY OF THE MODEL

OUTPUT VARIABLE -- SEEI INPUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFERENCES

SEEI RANDOM 1- 365

FARAMETER VARIABLE TYPE FACTOR ORDER ESTIMATE ST. ERR. T-RATIC 1 SEET AR 1 1 0.7856 0.0339 23.16 2 SEET AR 2 27 0.1039 0.0545 1.91

RESIDUAL SUM OF SQUARES = 826.987366 BEGREES OF FREEDOM = 305 RESIDUAL MEAN SQUARE = 2.498476

#### ( BACECASTS EXCLUDED )

ACF VARIABLE IS RESBEI. MAXDAC IS 60.7

FIRST CASE NUMBER TO BE USED = 29
LAST CASE NUMBER TO BE USED = 365
NO. OF OBS. AFTER DIFFERENCING = 337
MEAN OF THE (DIPPERNCED) SERIES = 0.2617
STANDARD BRBOR OF THE MEAN = 0.2653
T-VALUE OF MEAN (AGAINST ZERO) = 2.3652

## AUTOCORRELATIONS

1- 12 ST.E.			03 .06			
13- 24 ST.E.			0.0 .06			
25- 36 ST.E.		 	 .05 .06	 	 	 
37- 48 ST.E.			01			
49- 60 ST.E.			.02			

#### PLOT OF AUTOCORRELATIONS

-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0 LAG CORR, +---+---+---+ 1 0.238 + [XX+XXX 2 -0.155 X+XXI + 3 -0.124 XXXI ÷ 4 -0.172 X + X X I + 5 -0.097 + | | | | + + 11 -6 -0.027 7 -0.025 + 11 + 8 0.038 + IX + 9 0,019 · [] -0.051 + 12 -11 -0.021 + 7.7 15 0.000 4 IX •

13 0.022 + 11 + 14 -0.050 + XI + 15 0.031 0.018 16 -0.003 0.001 18 13 -0.083 20 -0.014 21 -0.002 22 -0.031 + XI 23 -0.018 + T 24 -0.025 + XI + 25 0.057 + IX + 26 0.056 + [] + 27 -0.021 + 11 + 28 0.007 + I 29 -0.006 30 0.054 31 -0.019 + I + 32 -0.022 + 11 + 33 0.021 + IX + 34 -0.028 + 11 + 35 -0.014 36 -0.033 + 11 + 37 -0.031 + XI + 38 0.039 + 11 + 39 -0.010 + I + 4 C 0.004 + I + 41 0.031 + II + 42 -0.027 + XI + 43 -0.030 44 -0.018 45 -0.007 + I 46 0.008 + I + 47 0.025 + [] + 48 -0.027 + 11 + 49 -0.058 + XI 50 -0.024 + 11 51 -0.022 + XI 52 -0.001 53 0.017 + I 54 0.020 55 0.067 + 1111+ 56 0.045 + 17 + 57 -0.029 + 11 + 58 -0.014 + I 59 -0.017 + I + 80 -0.024 + XI + EST

NUMBER OF INTEGER WORLS OF STORAGE USED IN PROCEDING PROCEDEY 7049 OPU TIME USED 22.780 SECONDS

BMEP2T - BOX-JENEINS TIME SERIES ANALYSIS

Mon Aug 4 17:02:42 1986

## BMDP Example Problem #2

Computer: CYBER

Data set used: Last two years of BECAT Model fit attempted: AR 1,27 MA 1

Good model

1PAGE 1 BMDP2T

BMDP2T - BOX-JENEINS TIME SERIES ANALYSIS
BMDP STATISTICAL SOFTWARE, INC.
1964 WESTWOOD BLVD. SUITE 202
(213) 475-5700
PROGRAM REVISED APRIL 1982
MANUAL REVISED -- 1981
COPYRIGHT (C) 1982 REGENTS OF UNIVERSITY OF CALIFORNIA

TO SEE REMARKS AND A SUMMARY OF NEW FEATURES FOR THIS PROGRAM, STATE NEWS. IN THE FRINT PARAGRAPH.

THIS VERSION OF BMDP HAS BEEN CONVERTED FOR USE ON CDC 6000 AND CYBER SERIES COMPUTERS BY BMDF PROJECT, VOGELBACK COMPUTING CENTER NORTHWESTERN UNIVERSITY 2129 SHERIDAN ROAD EVANSTON, ILLINOIS 60201 (312) 492-3681

RELEASED AUGUST 1983 FOR FINE COMPILERS

EXECUTED ON 86/10/11.AT 20.09.13.

PROGRAM CONTROL INFORMATION

/PROBLEM TITLE IS 'GEOSYNCH ENERGETIC ELECTRON DATA'.

/INPUT VARIABLES ARE 11.

FORMAT IS FREE.

RECLEN:97.

/VARIABLE NAMES ARE PERIOD, DATE, PTS, SEEIII,

GAMI, GAMIII, GAMIV, SEBI, SEBII.

SBRIV, SERSSO.

/TRANSFORM DELETE = 1 TO 730.

/END

PROBLEM TITLE IS

GEOSYNCH ENERGETIC ELECTRIN DATA

NUMBER OF CASES TO BEAD IN. . . . . . . . . . TO END MISSING VALUES CHECRED BEFORE OR AFTER TRANS. . NEITHER ELANES ARE. . . . . . . . . . . . . . . . . . MISEING REWINE INPUT UNIT PRICE TO READING. . DATA. . . NUMBER OF WORDS OF DYNAMIC STORAGE. . . . . . . 39998 \*\*\*\* TRAN PARAGRAPH IS USED \*\*\*\*\* 1PAGE 2 BMDP2T GBOSYNCH ENERGETIC ELECTRON DATA VARIABLES TO BE USED 1 PERIOD 2 DATE 3 PTS 4 SEEIII 5 GAMI 6 GAMIII 7 GAMIV 6 SEEL 9 SEEH 10 SEELV 11 SEESSD INPUT FORMAT IS PREE MAXIMUM LENGTH DATA RECORD IS 97 CHARACTERS. NUMBER OF CASES READ. . . . . . . . . . . . . . . . 1095 CASES WITH USE SET TO NEGATIVE VALUE . . . . 730 BEMAINING NUMBER OF CASES . . . . . . . . 365 1PAGE 3 BMDP2T GEOSYNCH ENERGETIC ELECTRON DATA VARIABLE IS SEBIII. ARIMA AROR = (1),(27). MAOR = '(1)'./THE COMPONENT HAS BEEN ADDED TO THE MODEL THE CURRENT MODEL HAS OUTPUT VARIABLE = SEEIII INPUT VARIABLE = NOISE 1PAGE 4 BMDP2T GEOSYNCH ENERGETIC ELECTRON DATA BSTIMATION RESIDUAL = RESERTIT./ ESTIMATION BY CONDITIONAL LEAST SQUARES METHOD RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN .1000E-04 SUMMARY OF THE MODEL

OUTPUT VARIABLE -- SEEIII INPUT VARIABLES -- NOISE VABIABLE VAR. TYPE MEAN TIME DIFFERENCES

SEEIII RANDOM 1- 365

 PARAMETER VARIABLE
 TYPE
 FACTOR
 ORDEE ESTIMATE
 ST. ERE.
 T-BATIC

 1
 SBEHII
 MA
 1
 1
 -1223
 .0852
 -2.21

 2
 SBEHII
 AR
 1
 1
 .9851
 .0096
 102.81

 3
 SBBHII
 AR
 2
 27
 .1694
 .0545
 3.11

RESIDUAL SUM OF SQUARES = .120532 DEGREES OF FREEDOM = .324 RESIDUAL MEAN SQUARE = .000391

ESTIMATION BY BACKCASTING METHOD

BELATIVE CHANGE IN BESIDUAL SUM OF SQUARES LESS THAN .1000E-04

SUMMARY OF THE MODEL

OUTPUT VARIABLE -- SEBIII INPUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFESENCES

SEETII RANDOM 1- 365

 PARAMETER VARIABLE
 TYPE
 FACTOR
 ORDER ESTIMATE
 ST. ERR.
 T-RATIO

 1
 SEEIII
 MA
 1
 1
 -.1227
 .0547
 -2.24

 2
 SEEIII
 AR
 1
 1
 .9850
 .0045
 218.90

 3
 SEEIII
 AR
 2
 27
 .1761
 .0543
 3.13

RESIDUAL SUM OF SQUARES = .130532 (BACECASTS EXCLUDED)

DEGREES OF FREEDOM = 334 RESIDUAL MEAN SQUARE = .000391

1PAGE 5 BMDP2T GBOSYNCH ENERGETIC BLECTRON DATA

```
ACF VARIABLE IS RESPELLE./
```

NUMBER OF OBSERVATIONS	:	355
MEAN OF THE (DIFFERENCES) SELIES	=	.5015
STANIARI ERROR OF THE MEAN	=	.0010
T-VALUE OF MEAN (AGAINST ZERC)	:	1.6901

## AUTOCCERELATIONS

1- 12	03	16	07	15	13	01	01	.07	0.0	.01	01	01
ST.E.	.05	.05	.05	.05	.08	.0€	.06	.06	.06	.06	.06	.00
13- 24	.12	06	01	01	0.0	04	.01	.01	.03	03	08	07
ST.E.	.06	.06	.06	.06	.06	.06	.06	.06	.06	.06	.06	.06
25- 36	08	.11	02	0.0	.04	01	07	.07	.05	0.3	04	08
ST.E.	.06	.06	.06	.06	.06	.06	.06	.06	.08	.06	.06	.06

## PLOT OF AUTOCORRELATIONS

	-1.08	64	2	.0	. 2	, 4	. £	. 6	1.0
LAG	CORE. ++-	+	+	- +	-+	+	-+	-+	- <del>-                                  </del>
				I					
į	026		4	XI +					
2	155		X + X	XI +					
3	075		÷ X						
4 5 6	151		X + X						
5	132		XX						
ĉ	014		+	I +					
7	011		+	•					
8	.975			IXX+					
9	005			Ţ +					
10	.011			I +					
11	024			X! +					
12	020			XI +					
13	.118			IIIX					
14 15	061			ΙΙ + Ι +					
15 16	005 011			I +					
19	002		† †	I					
18	040		•						
19	.009			I +					
<b>2</b> 0	.009			i +					
2]	.035		•						
20	015			¥I +					
23	673			XI +					
24	0^4			XI +					

25 .090 + []]+ 26 .114 + !!!!! 27 -.011 .004 29 .044 + [] + 30 -.015 31 -.071 + | | | | 32 .071 + []]. 33 .052 + IX + 34 .001 4 I 4 35 -.042 + XI + 36 -.084 + | | | | | | | 1PAGE 6 BHDP2T GEOSYNCH ENERGETIC BLECTRON DATA

FORECAST CASES = 24.

START = 345./

## FORECAST ON VARIABLE SEEIII FROM TIME PERIOD 345

PERIOD	FORECASTS	ST. ERE.	ACTUAL
348	.12734	.02037	.13540
34€	.12504	.03040	.10520
347	.12386	.0376€	.13230
348	.12352	.04356	.13000
349	.12151	.04881	.13500
350	.12201	.05305	.13790
351	.12453	.05702	.13420
352	.12073	.06063	.13500
353	.12124	.06394	.13600
354	.11836	.06699	.14050
355	.11581	.06982	.14090
356	.11300	.07247	.14240
357	.11136	.07494	.13649
358	.11018	.07727	.14210
359	.10759	.07946	.14190
360	.10763	.08153	.14210
361	.10566	.08349	.13349
362	.10449	.03535	.14239
3€3	.10270	.08711	.14740
3 € 4	.10177	.09879	.14483
365	.10008	.00003	.13530
3 € €	.03856	.09:9:	
3€7	.09882	.09686	
368	.03857	.03475	

STANDARI ERROR = .200740E-01 (BY CONDITIONAL METHOD )

IPAGE 7 BMCPCT GEOSYNCH ENERGETIC BLECTRON FATA

END '

NUMBER OF INTEGER WORDS OF STORAGE USED IN TRECEDING PROFILEM - ECOP-OPU TIME USED - 11.754 SECONDS 1PAGE - 8 BMDFCT

EMOP2T - BOX-JENKINS TIME SERIES ANALYSIS EXECUTED ON 86/10/11.AT 20.09.27.

PROGRAM CONTROL INFORMATION

NO MORE CONTROL LANGUAGE.

PROGRAM TERMINATED

#### BMOI Example Problem #0

Computer: CYBER Data set used: CCFLO Transfer Function Model

1PAGE 1 BMDP2T

EMDEST - BOX-JENKINS TIME SERIES ANALYSIS
BMDF STATISTICAL SOFTWARE, INC.
1964 WESTWOOD BLVD. SUITE 202
(213) 475-5700
PROGRAM REVISED APRIL 1982
MANUAL REVISED -- 1981
COFYRIGHT (C) 1982 REGENTS OF UNIVERSITY OF CALIFORNIA

TO SEE BEMARKS AND A SUMMARY OF NEW FEATURES FOR THIS PROGRAM, STATE NEWS. IN THE PEINT PARAGE PH.

THIS VERSION OF BMBP HAS BEEN CONVERTED FOR USE ON CEC 5000 AND CYBER SERIES COMPUTERS BY BMBP PROJECT, VOGELBACK COMPUTING CENTER NORTHWESTERN UNIVERSITY 2129 SHERIDAN RGAD EVANSTON, JULINOIS 60201 (2121 492-368)

RELEASED AUGUST 1983 FOR FTN5 COMPILEES

EXECUTED ON 86/10/19.AT 10.50.33.

#### PROGRAM CONTROL INFORMATION

/PROBLEM TITLE IS 'GEOSYNCH BE TIME SERIES,

TRANSFER FUNC FROM SOLAR WINE, IMF 32'.

/INPUT VARIABLES ARE 12.

FORMAT IS FREE.

RECL = 90.

/VARIABLE NAMES ARE DATE, ABXE, ABYE, ABZE,

ABYM, ABCM, AV, SEESSD, SEEI,

SEEII, SEEIII, SEEIV.

/TRANSFORM DELETE = 1 TO 506.

/ENI

PROBLEM TITLE IS

GENSYNOR BE TIME SERIES, TRANSFER FUNC FROM SOLAR WOW. DVE BO

NUMBER OF VARIABLES TO READ IN. . . . . . . . . . NUMBER OF VARIABLES ADDED BY TRANSFORMATIONS. . NUMBER OF CASES TO READ IN. . . . . . . . . . . TO ENI MISSING VALUES CHECKED BEFORE OR AFTER TRAKS. . NEITHER BLANKS ARE. . . . . . . . . . . . . . . . . . MISSING REWIND INPUT UNIT PRIOR TO BEADING. . DATA. . . NUMBER OF WORDS OF DYNAMIC STORAGE. . . . . . 49398 \*\*\*\* TRAN PARAGRAPH IS USED \*\*\*\*\* 1PAGE 2 BHDP2T GBOSYNCH EE TIME SERIES, TRANSFER FUNC FROM SOLAR WINE, IMP EL VARIABLES TO BE USED 1 DATE 2 APRE 3 ABYE 4 AESE 5 AEYM E AFZM 7 AV 8 SEESSI 9 SEEL 10 SEELI 11 SEBIII 12 SEEIV INFUT FORMAT IS FREE MAXIMUM LENGTH DATA RECORD IS 90 CHARACTERS. CASES WITH USE SET TO NEGATIVE VALUE . . . . IPAGE 3 BMDP2T GEGSYNCH BE TIME SERIES, TRANSFER FIND FROM SOLAR WINE, IMP ED ARIMA VARIABLE IS ABZM. ABOE = '(1)'. CENTERED./ THE COMPONENT HAS BEEN ADDED TO THE MODEL THE CURRENT MODEL HAS OUTPUT VARIABLE = ABZM INPUT VARIABLE = NOISE 1PAGE 4 BMDP2T GBOSYNCH BE TIME SERIES, TRANSFER FUND FROM SOLAR WIND, IMP ED ESTIMATION RESIDUAL = RX. METHOD IS CLS./ ESTIMATION BY CONTITIONAL LEAST SQUARES METHOL

RELATIVE THANGE IN BACH ESTIMATE LESS THAN 1,00000-00

STAMARY OF THE MODEL

CUTEUT VARIABLE -- ABOM INSUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFERENCES

ABOM RANIOM REMOVED 1- 200

PARAMETER VARIABLE TYPE FACTOR ORDER ESTIMATE ST. ESS. T-RATIC 1 ABBM AR 1 1 .7423 .0475 15.62

BESIDUAL SUM OF SQUABES = \$17.550939 BEGREES OF FREEDOM = 198 BESIDUAL MEAN SQUARE = 2.613894

1PAGE 5 BMEPST GEOSYNCH BE TIME SERIES, TRANSFER FUNC FROM SOLAR WINE, IMF 82

FILTER VARIABLE IS SEESSI.
RESIDUAL = RY./

BESIDUAL MEAN SQUARE = 45593.047731

VARIABLE SEESSD IS FILTERED, RESULTS ARE STORED IN VARIABLE BY 1PAGE & BMDP2T GEOSYNCH BE TIME SERIES, TRANSPER FUND BEIM SOLAR WIND. IMP ED

CCF VARIABLES ARE RX, RY. MAXLAG IS 10./

EFFECTIVE NUMBER OF CASES = 199

CORBELATION OF RY AND EY IS .CC

CROSS CORRELATIONS OF RX (I) AND RY (I+E)

TEASS CLERELATIONS OF BY (I' AND RY (I'E)

## TRANSFER PUNCTION WEIGHTS

	SCCF(X)	1,Y(I+E::	SCOFTY.	I),X(I-E'
LAG	*SY/SX	18X 88A	#SY/SX	18%.3%
Ċ.	.52041	.00003	.52041	.00000
•	-24.77278	00142	-5.77467	00003
2	8.57492	.00049	8.59414	.00649
3	-7.94904	00046	1.77053	.00010
4	12.56537	.00072	-6.46692	00037
5	14.97775	.00086	8.51527	.00049
6	7.12679	.00041	6.33562	.00036
?	11.63571	.00067	2.17164	.00012
8	-7.21447	00041	-1.13360	<b>0</b> 0000£
9	13.06466	.00075	-9.24338	00053
10	-3.49978	00020	4.87125	.00028

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD ERROR OF X(I), AND SY THE STANDARD ERROR OF Y(I)

1PAGE 7 BMDF2T GBOSYNCH BE TIME SERIES, TRANSFER FUNC FROM SOLAR WINE, IMP BZ

## PLOT OF CROSS CORRELATIONS

LAG CORE. +		-1.0	86	42 .	0 .2	. 4	.€	.8 1.3
-10 .037	LAG	CORE. +	+ + .			+	-+	-++
-8009	-10	.037						
-? .016		070		+ XXI	+			
-6 .048	-8	009		+ I	•			
-5 .064 + IXX +  -4049 + XI +  -3 .013 + I +  -2 .065 + IXX+  -1044 + XI +  0 .004 + I +  1188	- ?	.01€		+ I	+			
-4049	-8	.048		+ I	X •			
-3 .013		.064		+ I	XX +			
-2 .065		049		+ 11	÷			
-1044	- 3	.013		+ I	+			
0 .004 + I + I + I + I188	- 2	.065		+ I	<b>X</b> X+			
1188	-:	044		+ 11	+			
2 .065	0	.004		+ I	+			
3  060		188		<b>XX</b> + <b>X</b> X I	+			
# .00f	2	.065		+ I	X X -			
\$ .110		960		+ 77	•			
0 .054		.095		+ 1	77 -			
1 .056 + IXI + + IXI +		.:10		• !	XXX.			
€ -:555 + XI +	Ę			•	γ .			
	•	.014			X2 •			
9 .099 + 1X7 +	Ę	055		+ 7.7	•			
	ĝ	.093		4	X7 •			

-10 -.007 · XI +

1FAGE 8 EMOFOT GEOSYNCH BE TIME SERIES, TRANSFER FOND FROM SOLAR WINE. IMF ED

ARIMA VARIABLE IS SEESSI.

CENTEBES ./

THE COMPONENT HAS EBEN ADDED TO THE MODEL

THE CURRENT MODEL HAS OUTPUT VARIABLE = SEBSSD INPUT VARIABLE = NOISE

IPAGE 9 BMDP2T GEOSYNCH BE TIME SERIES, TRANSFER PUNC FROM SOLAR WIND, IMP EC

INDEF VARIABLE IS ABZM.

UPORDERS = '(1)'.

UPVALUES = -24.77278./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS
OUTPUT VARIABLE = SEESSD

INPUT VARIABLE = NOISE ABZM

IPAGE 10 BHCP2T GEOSYNCH BE TIME SERIES, TRANSFER FUNC FROM SCLAR WIND, IMF 82

ESTIMATION RESIDUALS = BYX.
METHOD IS CLS./

ESTIMATION BY CONDITIONAL LEAST SQUARES METHOD

RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN . 11908-04

SUMMARY OF THE MODEL

CUTPUT VARIABLE -- SEESSD

INPUT VARIABLES -- NOISE ABZM

VARIABLE VAE. TYPE MEAN TIME DIFFERENCES

SEESSI RANDOM REMOVED 1- 200

ABBM RANDOM 1- 200

```
PARAMETER VARIABLE TYPE FACTOR CEDER ESTIMATE SOLEROL THOUSE
   1 AB2M UP 1 1 -8.801 7.0040 -1.10
RESIDUAL SUM OF SQUARES = 14710279.871337
DEGREES OF FREEDOM
RESIDUAL MEAN SQUARE = 74294.342785
IPAGE II BHDP2T GEOSYNCH EE TIME SERIES, TRANSFER FUNC FROM SOLAE WIND, IMP BO
ACF
         VARIABLE IS RYX./
NUMBER OF OBSERVATIONS =
MEAN OF THE (DIFFERENCED) SERIES = 4.9694
STANDARD BRBOR OF THE MEAN = 19.3187
T-VALUE OF MEAN (AGAINST ZERO) = .2572
AUTOCORRELATIONS
 1- 12 .65 .42 .20 .12 .01 -.09 -.10 -.12 -.02 .02 .06 .04
        ST.E.
13- 24 .06 .05 -.03 -.07 -.10 -.09 -.12 -.04 -.01 0.0 .05 .08
 ST.E.
        $2. (0. 30. 30. 80. 11. E1. 22. 32. 82. 10. 71. 36 -32
 ST.E. .11 .11 .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13
PLOT OF AUTOCCRRELATIONS
      -1.0 -.9 -.6 -.4 -.2 .0 .2 .4 .6 .8 1.0
LAG CORE. +---+---+---+---+----+----
 1 .646
                         + IXX+XXXXXXXXXXXXX
 2 .423
                         + IXXXX+XXXXX
     .201
                         + IXXXXX
     .122
                         + IXXX +
     .013
 6 -.091
                        + %%I
 7 -.104
                        + XXXI
    -.120
                         + XXXI
 9 -.019
                         + I
 10
     .024
                            .062
                         + ! ! ! ! -
```

+ ! X

- IX -

÷ 7.

+ XXI

137 •

.042

.061

.054

15 -.031

16 -.070

```
17 -.100
                            + XXX!
18 -.092
                            + XXI
    -.117
10
                             + XXX
    -.041
                             . 71
20
    -.010
22
     .001
                                  IX
23
     .050
     .083
24
                                 IXX +
25
    .165
                                 I X X X X +
     . 227
26
                                 IXXXXXX
27
     .273
                                I X X X X X + X
28
     .282
                                 IXXXXX+X
29
     .221
                                 IXXXXXX
     .185
30
                                 IXXXXX+
31
     .120
                                 IXXX +
32
     .077
                                 IXX +
33
     .061
                                 IXX
34
     .059
                                 ŢΧ
35
     .062
                                 IXX
36
     .041
                                 IX
```

1PAGE 12 BMDP2T GEOSYNCH EE TIME SERIES, TRANSFER FUNC FROM SQLAR WIND, IMF BZ

## PACF VARIABLE IS RYX./

NUMBER OF OBSERVATIONS = 199
MEAN OF THE (DIFFERENCES) SERIES = 4.9694
STANDARD ERROR OF THE MEAN = 19.3187
T-VALUE OF MEAN (AGAINST ZERO) = .2572

## PARTIAL AUTOCORRELATIONS

#### PLOT OF PARTIAL AUTOCORNELATIONS

```
-1.0 -.8 -.6 -.4 -.2 .6 .2 .4 .6 .7 1.0
: .648
                         2 .009
                         + 1 +
 3 -.130
                          XXXI +
                         + IXX+
   .088
 5 -.094
                          + | | | | +
 6 -.120
                         XXX : +
 7
   .059
                          + [] +
   -.048
 8
                          + 11 +
9
   .136
                          + []]]
10
   .020
11 -.012
                          + I +
12 -.027
                          + 11 +
   .036
13
                          + IX +
14 -.020
                          + I +
15 -.123
                          XXXI +
15
    .019
                          + I +
   .000
17
                          + I +
18 -.025
                          + 11 +
19 -.042
                          + 11 +
20
    .112
                          + 1222
   -.017
   -.052
                          + XI +
23
   .086
                          + []]+
24
   .021
                          + IX -
   .121
25
                          + IXXX
26
   .167
                          + [][-1
2"
                          + [] +
    .031
28
   .092
                          + [XX+
29 -.016
30 .011
31 -.002
32 .025
33
   .095
   .008
34
35
   .030
                          + 11 +
38 -.011
                          + 1 +
```

1PAGE 13 BMEP2T GEOSYNCH BE TIME SERIES, TRANSPER FUNC FROM SCLAR WIND, IMP B2

ARIMA VARIABLE IS SEESSE.

CENTERED.

ABOR = '(1,27]'./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS

GUTFUT VARIABLE = SEESSI

INPUT VARIABLE = NOISE ABZM

IPAGE 14 BHDPCT GEOSYNCH BE TIME SERIES, TRANSFER PUNC FROM SOLAR WIND. IMP 52

ESTIMATION RESIDUALS = RYX./

ESTIMATION BY CONDITIONAL LEAST SQUARES METHOD

RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN 11000E-04

SUMMARY OF THE MODEL

OUTFUT VARIABLE -- SEESSD

INPUT VARIABLES -- NCISE ABZM

VARIABLE VAR. TYPE MEAN TIME DIPPERENCES

SEESSE RANDOM REMOVED 1- 200

ABOM RANDOM 1- 200

PARAMETER VARIABLE TYPE FACTOR ORDER ESTIMATE ST. BES. T-RATIC 1 SEESSD AR 1 1 .7033 .0524 13.42

2 SEESSD AR 1 27 .1339 .0528 2.63 3 ABZM UP 1 1 -15.83 8.1434 -1.94

RESIDUAL SUM OF SQUARES = 5434016.344292 DEGREES OF FREEDOM = 169 RESIDUAL MEAN SQUARE = 32155.718014

ESTIMATION BY BACKCASTING METHOD

RELATIVE CHANGE IN RESIDUAL SUM OF SQUARES LESS THAN .1000E-04

SUMMARY OF THE MODEL

OUTPUT VARIABLE -- SEESSO

INPUT VARIABLES -- NOISE ABZM

VARIABLE VAR. TYPE MEAN TIME CIFFERENCES

SEESST RANDOM BEMOVED :- CCC

ABOM RANCOM 1- 201

```
PARAMETER VARIABLE TYPE FACTOR ORIER ESTIMATE
                                        ST. ERR. 7-RATI
                     1
                                        .0479
               AE.
                                                 12.78
    : SEESSI
                                  .6115
                                  .1950
    2 SEESSI
               AE.
                                          .0481
                                                  4.09
                      1 -24.01
                UP
                                        7.0927 -3.10
    3 APOM
RESELVAL SUM OF SWIARES : 5578890.E98109 (EACHMASTS ENGLICED)
DEGREES OF FREETOM =
                        169
RESIDUAL MEAN SQUARE = 000011.199794
IPAGE 15 BMOPOT GEOSYNOH EE TIME SERIES, TRANSPER FUNC FROM SOLAR WIND, IMP BO
```

ACF VARIABLE IS RYX.
MAXLAG IS 36./

NUMBER OF OBSERVATIONS = 200
MBAN OF THE (DIFFERENCED) SERIES = -.8012
STANDARD ERROR OF THE MEAN = 14.1529
T-VALUE OF MEAN (AGAINST ZERO) = -.0566

### AUTOCORRELATIONS

	-1.08	842	.0	. 2	. 4	.6	.9 1.0
DAC	CORE. ++	-+++	<del>-                                    </del>	-+	- +	-+	-++
			:				
:	013	÷	Ţ. +				
0	.085	•	IIX+				
3	104	XX	XI +				
4	.055	+	IXX	÷			
5	001	+	Ī	÷			
6	113	+ <b>X</b> X	ΧI	+			
7	043	+	XI	+			
8	135	÷ X X	ΥĪ	÷			
õ	.062	÷	IXX	<del>ļ</del>			
10	.623	+	Ţ	4			
1:	.112	÷	IXXX	•			
	.018	÷					
:3	.852	+	Ţ Ÿ	_			
• •	.105	+	IXXX				
15	056	•	ΣÏ				

18 -.(22 · Y: · -.081 + 111 -18 800. + : 19 -.090 • 111 + [] + .031 .075 · 111 · 20 -.012 23 .051 + IX + 24 -.006 25 .035 + 15 + 2€ .088 + IXX + 2? -.088 + XXI + 28 .034 + 17 + 29 -.043 + 1 30 .065 + IXX + 31 -.001 32 -.004 33 .016 34 .042 + II + 35 .102 + IXXX+ 3€ .025 II ÷

IPAGE 16 BHDP2T GEOSYNCH BE TIME SERIES, TRANSFER FUNC FROM SOLAR WIND, IMF EZ

ARIMA VARIABLE IS ABOM.

AROR = '(1)'.

ARVALUES = 0.7423.

CENTERED./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

24 PSI-WEIGHTS ARE STORED.

1PAGE 18 BMDP2T GROSYNCE BE TIME SERIES, TRANSFER FUNC FROM SOLAR WIND, IMF BE

BRASE MODEL./

UNIVARIATE TIME SERIES MODEL ERASED IFACE 19 EMEPAT GEORYNCH SE TIME SERIES, TRAMSFER PONC FROM SOLAR WOND, IMPOR ARIMA VARIABLE IS SEESST.

AR'E = "(1,27)".

ABVALUES = 0.8125, 0.1970.

CENTERED.

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS OUTPUT VARIABLE = SEESST INPUT VARIABLE = NOISE

1PAGE 20 BHDF2T GEOSYNCH EE TIME SERIES, TEANSFER FUNC FROM SCLAR WIND, IMF B2

INDEP VARIABLE IS AREM.

UPORDERS = '(1)'.
UPVALUES = -24.01.
CENTERED./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS

OUTPUT VARIABLE = SEESSD

INPUT VARIABLE = NOISE ABZM

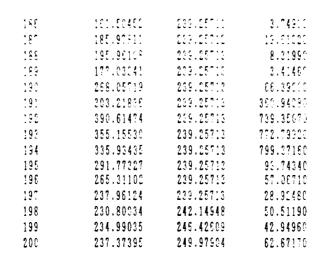
IPAGE 21 EMOP2T GEOSYNCH BE TIME SERIES, TRANSFER FUNC FROM SOLAR WIND, IMT BO

FORECAST CASES = 30.

START = 171./

FORECAST ON	VARIABLE	SBBSSD	FROM	TIME	PERICE	171

PERIOD	FORECASTS	ST. BRR.	ACTUAL
171	E7.28424	189.12573	24.40170
172	269.91766	221.78217	8.91490
173	214.72861	232.85505	12.55530
174	190.19581	236.87563	29.57889
175	173.39074	238.36649	11.77450
17€	20€.66543	238.92339	12.85010
177	204.17370	239.13198	17.24380
178	111.97103	239.21619	16,15370
11.9	144.80842	239,23952	7.31580
180	125.66976	239.25053	€.00410
181	154.95765	239.25465	4.38110
182	157.36364	239.25820	3.97570
186	152.44244	239.25678	4.38460
184	164.83170	109.25700	3,02040
188	240,25144	209.25708	3.[9[9(



STANDARD BERGE = 189.126 (BY CONDITIONAL METHOD);
1PAGE 22 BMDP2T GEOSYNCH BE TIME SERIES, TRANSPER FUNC FROM SOLAR WIND, IMP ED

BND/

NUMBER OF INTEGER WORDS OF STORAGE USED IN PRECEDING PROFILEM 5988 CPU TIME USED 9.070 SECONDS

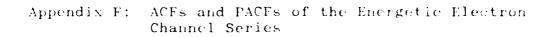
IPAGE 23 EMERCE

EMEPST - BOX-JENKINS TIME SERIES ANALYSIS EXECUTED ON 86/10/19.AT 10.50.47.

PROGRAM CONTROL INFORMATION

NO MORE CONTROL LANGUAGE.

PROGRAM TERMINATED



							J	Puge
ACF of the SEESSD Channel, 706 days	•					•	•	164
PACF of the SEESSD Channel, 706 days		•						160
ACF of the SEEI Channel, 765 days .	•						•	168
PACF of the SEEI Channel, 765 days	•							170
ACF of the SEEII Channel, 765 days	•		•		•		٠	172
PACF of the SEEII Channel, 765 days	•		•		•		•	174
ACF of the SEEIII Channel, 765 days	•			•			•	176
PACF of the SEEIII Channel, 765 days							•	178
ACF of the SEEIV Channel, 765 days		•		•	•			180
PACF of the SEEIV Channel, 765 days								182

Appendix F: ACFs and PACFs of the Energetic Electron Channel Series

#### ACE VARIABLE IS SIESSI

NUMBER OF OBSERVATIONS	:	506
MEAN OF THE (DIFFERENCELL SERIES	:	208,8009
STANDARD BRROR OF THE MEAN	:	8,8395
T-VALUE OF MEAN (AGAINST ZERO)	=	23.2595

#### AUTOCORRELATIONS

1- 12 ST.E.								30 30.				
13- 24												
ST.E.	.06	30.	.05	.0E	.06	.oe	.05	.05	.06	.06	.05	.01
25- 36	.22	.28	.35	.26	. 23	.18	.13	.10	. ( ?	.06	.00	1.0
ST.E.	.06	.05	.0€	.Of	.0€	.07	.07	.07	.07	.07	.07	.07

#### PLOT OF AUTOCOBRELATIONS

LAG		86								
LAS	CCAA. ******	*****			ţ		••••	• • • • •	•••	
1	.594			÷		IIII	XXXXX	XXX		
	.416			Ļ	17+	XXXXX	ZΣ			
63 63	.248					+ 7 7 X				
4	.124		•		IXX					
5	.085				IXX					
6	.018				Ĭ					
?	012		+		Ţ					
8	055		÷	X		+				
ş	004		+		Ţ	÷				
10	.0:2		•		Ī	•				
11	.039		•		[ <u>Y</u>	4				
12	.050		+		IX					
	.050		÷		IX					
14	.067		+		122	•				
15	.978		+		IXZ	•				
16	.043				12	+				
: *	.022		+		ΙX	+				
18	010		+		I					
19	041		+	X	:	+				
20	.018		+		I	÷				
0.1	.022		+		] X	•				
20	.040		•		ΙX	+				
23	.078		+		IXX					
24	.148		+		IXX	<b>+</b> X				
25	. 201		+		IXX	<b>+XXX</b>				
2€	.282		+		IXX	+ 2 2 2 2				
2?	.316		+		IXX	+ X X X X	Y			
23	.260		ŧ		IXX	+ X X X				
29	.231		+			+ X X X				
<b>3</b> 0	.180		•		IXX	+ <u>X</u> X				
31	.131		÷		IXX					
35	.104		+		[ <b>X</b> X					
33	.972		+		IXX					
3 4	.085		+		ŢXX					
35	.005		+		1 1					
36	000		4		Ţ _	•				

## PARE VARIABLE 18 SIESSI

NUMBER OF CESERVATIONS	:	• .
Makk F THE COLFELBANCED SPRIES	=	277 8727
STANTART BRRUS OF THE MBAN	2	3.8035
CHVALUE OF MEAN MACAINST DEBOT	:	23,2537

:- ! <u>2</u> \$7.E.						
13- 24 ST.E.						
25 - 38 97 R						

## PLOT OF PARTIAL AUTOCCLEFIATIONS

1 4 7	-1.09642	.0 .4 .6 .9 .1.1
284	L. O. Sedie 1	:
1	.594	+ IX+XXXXXXXXXXXXX
0	. 999	- IXX
3	052	+XI +
4	050	+XI +
5	.040	+ IX+
6	049	+XI +
7	018	+ I +
8	047	+XI +
ç	.092	+ IXX
10	.003	+ [ +
11	.024	+ IX+
12	.010	+ 1 +
. 5	.9.7	+ I +
14	.017	+ : +
15	.027	+ IX+
16	040	+X: +
17	002	÷ 🛴 +
19	027	+XI +
19	001	*XI +
20	.101	+ IX+X
2:	.019	+ I +
9.2	003	1+ 1 +
25	.000	+ 188
24	.100	+ IX+X
25	.114	+ IX+X
25	.994	+ ! XX
27	, 089	+ IXX
28	007	+ I +
29	.036	+ IX÷
30	.011	+ I +
31	.010	+ I +
32	.030	+ IX÷
33	.027	+ IX÷
34	.034	+ IX+
3.5	041	+ X I +
3€	942	+XI +

## ACE VARIABLE IS SEE!

NUMBER OF OPSELVATIONS	Ξ	765
MEAN OF THE (LIFFEBENIED: SERIES	:	1,1,7
STANDARD BRROK OF THE MEAS	Ξ	.133
T-VALUE OF MEAN (ACAINST ZBRO,	:	12.7001

## AUTOCORRELATIONS

1- 12 ST.E.							
12- 24 ST.E.	.03					.0: .0£	
25- 06							
ST.E.						 	

## ACE VARIABLE IS SEE!

## PLOT OF AUTOCORRELATIONS

1	LAC	-1.5 -,3	842 .1 .2 .4 .5 .8 1.5
2	,.hu	Communication	
2	1	.733	+ IX+XXXXXXXXXXXXXX
2	2		+
5       .012       + I + I + I + I + I + I + I + I + I + I		.245	+ IXX+XXX
6017		.095	+ IXX+
7021 8019 9010 11	5	.012	<b>+</b> ↓ +
3	6	017	+ I +
9010 10 .015 11 .013 12 .017 13 .023 14	7	021	+ XI +
10	9	019	+ I +
11	9	010	
10	10	.015	
13 .033		.313	
14	12	.017	
15	13	.033	+ IX +
18 .099		.041	+ IX +
10 .078		.073	÷
18 .013			• IXX+
19026 20028 21033 22039 23 .010 24 .084 25 .151 26 .208 27 .252 28 .221 29 .171 30 .112 31 .023 31 .023 32025 33034 34040 25050 4 XI + 21 + 22050 4 XI + 23050 4 XI + 24040 25050 4 XI + 25050 4 XI + 26050 4 XI + 27050 4 XI + 28050 4 XI + 29 .171 4 XX			÷ IXX+
20028	: :		
21033			
20			
28 .010			
24 .084	55		
25			
26 .208			
27 .252			
28 .221			
29 .171			
30 .112 + IXXX 31 .023 + IX + 32025 + XI + 33034 + XI + 34045 + XI + 25050 + XI +			
31 .023 + IX + 32025 + XI + 33034 + XI + 34040 + XI + 25050 + XI +			
32025 + XI + 33034 + XI + 34045 + XI + 25050 + XI +			
30034 + XI + 34040 + XI + 25050 + XI +			
34040			
35 050 + XI +			
36009 + XI +			
	3.5	029	+ XI +

#### PACE VARIABLE IS SEE!

NUMBER OF OBSERVATIONS	:	78.5
MEAN OF THE (DIFFERENCE) SERIES	=	1.4100
STANDARD BRROK OF THE MEAN	Ξ	.1099
T-VALUE OF MEAN (AGAINST ZERI)	:	10.7009

1- 12 ST.E.							
13- 24 ST.B.	0.C .04						
25-36 ST.E.						CC 10	

## PACE VARIABLE IS SHIT

## PLOT OF PARTIAL AUTOCHBRELATIONS

LAG		0.1 8. 0. 0. 4. 0. 0. 0. 0. 4. 0. 8
ыкс	ccan.	<u>`</u>
;	.533	• IX-XXXXXXXXXXXXXXXXXXXXX
2	243	XXXX+XI +
3	.038	+ IXX
4	146	XX+XI +
5	.059	+ IX+
€	024	<b>+X</b> [ +
7	.02€	<b>+ IX+</b>
8	023	<b>+</b> X! +
9	.020	+ I +
10	.030	• IX÷
11	043	+ <b>X</b> I +
12	.048	<b>+</b> IΣ +
13	.005	+ I +
14	.022	+ IX+
15	.058	+ IXX
16	.012	+ I +
17	042	+∑; +
18	031	XXI ÷
19	.029	+ IX+
20	.013	<b>→</b> [∑-
21	034	÷X. +
22	.013	+ I +
23	.075	+ IXI
24	.092	• IXX
25	.058	• [X+
26	.078	+ IXX
2?	.075	+ IXX
2.8	058	*X. •
29	.047	+ II+
30	057	+Σ! +
31	057	+17 +
32	.021	+ IX+
33	.005	+ : +
34	.001	• <u>I</u> •
35	025	+XI_+
36	.023	* <u>!</u> I.*

## ACE VARIABLE IS SEE!!

NUMBER OF OBSERVATIONS	:	781
MEAN OF THE (DIFFERENCED) SERIES	:	.1512
STANDARD ERROR OF THE MEAN	:	.0087
T-VALUE OF MEAN (AGAINST ZERO)	=	17.5638

#### AUTOCORRELATIONS

:- !2 ST.E.												
13- 24 ST.E.												
25 - 36	.18	. 24	. 28	.24	.20	. 14	.06	.02	.01	01	- 01	-,:4

## ACE VALUABLE IS SIBIL

## PLOT OF AUTOCORRES ATIONS

-1.0	c	С		•	r	•		•	0	•
	• •		• 1				, -	• •		

LAC	-1.08 CORR. ++	
LEG	Cons. Torrette	Ţ
1	.705	+ 17+88888888888888
2	.435	+ 1XX+XXXXXXXX
3	.299	+ IXX+XXXX
4	.142	+ IXX+X
5	.080	+ IX +
6	.033	+ IX +
7	.010	+ I +
8	008	+ I +
9	011	+ I +
10	.008	+ I +
11	.000	÷ I +
12	005	+ I +
13	.011	+ I +
14	.012	+ I +
15	.029	+ IX +
16	.046	+ IX +
17	.041	+ IX +
18	.003	+ I +
19	022	+ XI +
20	005	+ <u>T</u> +
21	003	+ I +
22	.014	+ I +
23	.085	+ IXX+
24	.127	+ IXXX
25	.180	+ IXX+X
26	.242	• IXX+XXX
27	.282	+ IXX + XXXX
28	.238	÷ IXX+XXX
29	.19€	+ IXX+XX
30	.145	+ IXX+X
31	.061	+ IXX+
32	.020	+ I +
33	.005	+ I +
34	015	+ I +
35	637	<b>+ X</b> I
3 €	02€	+ XI +

## PACE VARIABLE IS SEE!!

NUMBER OF OBSERVATIONS	=	765
MEAN OF THE (DIFFERENCED) SERIES	:	.1500
STANDARD BRROE OF THE MEAN	=	3:10.
T-VALUE OF MEAN (AGAINST ZERO)	:	15,5008

1- 12 ST.E.											04 .04	
13- 24	. 02	0.0	.04	0.0	0.0	07	.01	.04	01	.04	en.	.:0
ST.E.												
25- 3€	.08	.11	.05	05	.04	04	04	.61	0.0	0.0	04	· •
ST T												

## PACE VARIABLE IS SEELI

## PLOT OF PARTIAL AUTHOCERELATIONS

LAG	CORR. +	
1	.705	+ !X->XXXXXXXXXXXXXXXX
2	125	X+XI •
3	383.	+ IXX
4	-,153	XX+XI +
5	.058	+ 17+
6	006	+ I +
?	.008	+ I +
8	025	+ X [ +
9	.009	† I ÷
10	.036	+ IX+
11	039	+ 🛽 [ →
12	.011	+ I +
13	.018	+ I +
14	004	<b>†</b>
15	.044	+ IX+
16	.001	+ I +
17	004	+ I +
18	067	XXI +
19	.007	+ T +
29	.045	+ ፲፮÷
21	011	÷ I ÷
22	.040	÷
23	.061	<b>+</b> Ι <b>Χ</b> Σ
24	.098	+ 188
25	.063	+ ! <b>X</b> X
26	.109	₹ \(\bar{\chi}\)\(\bar{\chi}\)
27	.061	+ IXX
28	047	÷X
29	.041	+ IX-
<b>3</b> C	039	+X. +
3 !	041	+ <b>X</b> I ÷
32	.012	+ I +
33	.002	+ I +
34	004	+ I +
35	036	+XI +
36	.018	+ 1 ÷

## ACF VARIABLE IS SEEIII.

NUMBER OF OBSERVATIONS	:	765
MEAN OF THE (DIFFERENCED) SERIES	:	.2560
STANDARD ERROR OF THE MEAN	:	.0412
T-VALUE OF MEAN (AGAINST ZESO)	:	€.4255

#### AUTOCORRELATIONS

1- 12 ST.E.		.21					
13- 24 ST.E.							
25- 36 ST.E.							

## ACE VARIABLE IS SEEIII

## PLOT OF AUTOCORRELATIONS

LAC		3 3	42		2. 0.	, !	. €	. 3	
FUA	Q-2-3838-4 - 1 - 1				•			,	
:	.535			4	IX-XXIX	XXXXXXX			
	. 3 ! 6			÷	IX-XXXX	ΣX			
5 5	.208			÷	IX+XXX				
4	.097			4	IXX				
5	012			÷	I t				
6	011			+	I +				
?	010			÷	I+				
3	010			÷	Ţ +				
g	010			ŧ	I +				
10	011			÷	Ţ +				
11	012			÷	I +				
12	013				I +				
13	014				I +				
14	.097			ŧ	IXX				
15	.208			÷	IX+XXX				
16	.209			t	I X + X X X				
17	.210			÷	I X X + X X				
18	.210			÷	X				
19	.100				IXXX				
20	010			÷	ī +				
21	011			÷	I +				
22	013			÷	I +				
23	014			÷	: •				
	015			٠	<u>.</u> +				
25	016			4	I 4				
56	016			-	: •				
27	017			+	I •				
23	.107			•	IXXX				
29	014			Ť	I t				
3€	014			÷	Ţ •				
0.	013			÷	i •				
22	012			÷	I +				
30	011			÷	I +				
34.	010			4	I +				
3.5	0:0			÷	I +				
3€	010			÷	; +				

## PACE VARIABLE IS SELLE.

NUMBER OF GESERVATIONS	:	195
MEAN OF THE (DIFFERENCED SPRIES	:	.5591
STANIARS BRECK OF THE MEAN	:	.0402
T-VALUE OF MEAN (AGAINST ZERC)	Ξ	8.4088

1- 12 ST.E.						
13- 24 ST.E.						
25- 36 ST.E.						

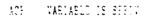
#### FACE VARIABLE IS SEELII

#### PLOT OF PARTIAL AUTOCORNELATIONS

<b>%</b>			FACE VARIABLE IS SEELLI
			FLOT OF PARTIAL AUTOCORNELATIONS
	112		1 3. 3. 4. 2. 9. 2 4 3
	LAS 1	.535	: + IX+XXXXXXXXXXX
	2	.042	+ IX+
	; 4	.029 045	+ X
	9 5	040 084	*A XX! +
	6	.038	<b>+ IΣ</b> +
	7	.005	* X *
	9 9	.006 007	+ I + + I +
	10	013	• I •
	11	062	+ I +
	12 13	004 003	+ I + + I +
	14	063 .153	+ IX+XX
	15	. 153	+ IX+XX
	16	.025	+ IX+ + IX+
;	15 18	.042 .040	† 1A+ † IX+
	19	077	XXI +
	20	075	XXI +
	21 22	.03£ .019	+ ΙΣ- + Ι +
	0.0 LL	.018	† I ÷
	24	018	* <u>T</u> +
	25 26	624 662	±XI • + I +
	25 25	.012	• I •
	28	.200	+ IX+XXX
	29	210	XXX-XI +
	30 3:	003 048	* · · · · · · · · · · · · · · · · · · ·
	2.2	949	•X: •
	33	007	4 ₹ 4 
	3. <b>4</b> 2.5	012 .080	· [ •
	36	009	* * *
<b>77.</b> 3			
			179
			1 ( 7

A TIME SERIES ANALYSIS OF EMERGETIC ELECTRON FLUXES (12 - 16 MEY) AT GEOS (U) AIR FORCE INST OF TECH WRIGHT-PATTERSON AFB OH SCHOOL OF ENGI M P HALPEN DEC 86 AFIT/G50/ENS-ENP/860-1 F/G 22/1 76-0194 360 UNCLASSIFIED NL





NUMBER OF OBSERVATIONS	:	768
MEAN OF THE (DIFFERENCED) SERIES	:	1.7:00
STANIARD ERROR OF THE MEAN	:	.3841
T-VALUE OF MEAN (AGAINST ZEEC)	:	1.7380

#### AUTOCORRELATIONS

1- 12 ST.B.						0.0	
13- 24 ST.E.	0.0					0.0 .04	
25- 36 ST.E.							

## ACF VARIABLE IS SEELV

## FLOT OF AUTOCORRELATIONS

	-1.0	8 -	.e4	2	.0	. 2	, 2	.€	.8	1.0
LAG	CORR	•	· + <del>†</del>	•	- <del>† ·</del>	+	· • • • • •	- +	-+	<del>†</del>
1	002				I +					
2	002				I +					
3	002				I +					
4	003			+						
£	003			+						
6	003			+	Ī +					
?	003			+	I +					
8	003			+	I ÷					
9	000			+	Ι÷					
10	00:			ŧ	I +					
11	003				I •					
12	003				I +					
12	003				I +					
14	003				Ţ.+					
15	003				Ţ÷					
16	003				I +					
17	003				[ ÷					
18	003				I +					
19	003				I +					
20	003				Ţ +					
21	002				I +					
22	003				I +					
23	003				I +					
24	003				I +					
25 ac	002				· 1 +					
26 27	002 003				· I +					
28	003				· I +					
20 29	003				· [ +					
30	003				Ţŧ					
31	003				. <u></u> . <u></u> .					
32	003				·					
33	003				· ː ·					
34	003									
35	,199				-   <u> </u>	Y 7 Y				
3 €	503				· [ +					
					-					

## PACE VARIANPE IS SEED ...

NUMBER OF OBSERVATIONS	:	785
MEAN OF THE (DIFFERENCED) SERIES	:	• • • • • •
STANDARY BERGR OF THE MEAN	:	.9840
T-VALUE OF MEAN (AGAINST ZERO)	:	1.5080

1- 12 ST.E.						0.3 .04	
13- 24 ST.E.						0.0	
25- 36 ST.E.						.20	

## PACE VARIABLE IS SEED.

## FIGT OF PARTIAL AUTOCORRELATIONS

		- 8 -	.E42 +	.0	. 2	. 4	, €	. 6
LAG	00hh. +-		·+-	+ I		+	+	+
1	002			+ I +				
2	001			+ [ +				
3	003			• I +				
4	003			+ 1 +				
5	003			+ [ +				
6	003			+ I +				
7	001			+ I +				
8	003			+ I +				
9	003			+ 1 +				
10	003			+ [ +				
11	003			+ 1 +				
12	003			+ [ +				
10	003			+ I +				
14	003			+ I +				
15	003			+ I +				
16	003			+ i +				
17	003			+ I +				
18	003			+ I +				
19	003			+ I +				
20	003			+ [ +				
21	001			+ I +				
22	003			+ I +				
23	003			+ [ +				
24	003			+ I +				
25	003			+ I +				
2€	003			+ I +				
27	003			+ I +				
28	003			+ I +				
29	001			+ I +				
30	003			+ I +				
31	003			+ [ +				
32	003			+ I +				
33	003			+ I +				
34	003			+ [ -				
3 £	.199			+ 11				
3 E	001			+ 1 4				

# Appendix G: ACFs and PACFs of the IMF $B_{\mbox{\scriptsize z}}$ and the Solar Wind Series

AND A CONTRACT OF SECTION OF SECTIONS OF SECTION OF SEC

								Page
ACF of the IMF $B_z$ Series						•	•	185
PACF of the IMF $B_z$ Series	 •	•			•	•	•	187
ACF of the Solar Wind Series .		•	•			•		189
PACF of the Solar Wind Series								191

## Appendix G: ACFs and PACFs of the IMF B, and the Solar Wind Series

ACF	٤,	• 0	7	4 5	٠,	T	٠.	r.	
A . *	٧.			٠,	•		•	۰	
11.6	•		•	***		•	 -	٠	·

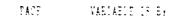
NUMBER OF GESERVATIONS	:	706
MEAN OF THE (DIFFERENCED) SERIES	:	.3811
STANDARE BRECK OF THE MEAN	:	1880.
T-VALUE OF MEAN (AGAINST ZERO)	:	4.5841

#### AUTOCORRELATIONS

. 02	.00			.00	. 0 0	.06	. 00
16	20	(1		ηŋ	0.4	٥E	0.0
.06	.06	.06	.56	.06	.06	. 96	.08
.10	.11	.12	82.	.02	01	01	01
	.06 10 .06	.06 .06 1008 .06 .06 .10 .11	.06 .06 .06 100804 .06 .06 .06 .10 .11 .12	.06 .06 .06 .00 .00 .00 .01 .01 .01 .02 .06 .06 .06 .06 .00 .00 .00 .00 .00 .00	.06 .06 .06 .06 .06 .06 1008040102 .06 .06 .06 .06 .06 .06 .10 .11 .12 .08 .02	.06 .06 .C6 .06 .06 .06 1008040102 .04 .06 .06 .06 .06 .06 .06 .06 .10 .11 .12 .08 .0201	.17 .09 .05 .05 0.00711 .06 .06 .06 .06 .06 .06 .06 .06 .06 .06

## PLOT OF AUTOCORRELATIONS

		88 -++	4	2 .3	.2	. 4	. €	. 2	1.5
LAG	CORK. *	<del>+</del> +-	+			•	•	• • • • •	•••
1	.633			-	+ X X X X X	X	XXX		
2	.414			+ IX	•XXXXX	ΣΧΣ			
3	.296			1 1 X	X - X X X X				
4	.227			• II	X - X X X				
5	.17:			+ 17	X • Z				
6	.089			+ []	<u> </u>				
7	.057			+ []	•				
8	.047			+ 11	+				
9	.003			+ I	+				
10	089			+ 🛚 🗸 🗸 T	÷				
11	113			XXXI					
12	142			X+ZZI					
13	154			<b>X+XX</b> I					
14	137			XXXI					
15	112			XXXI					
16	039			+ \ \ \					
17	104			III.					
18	077			+ % % [					
	037			+ [[					
20	013			• I					
21	017			+ I	ŧ				
22	.035			+ []	<b>(</b> +				
23	.046			+ 1					
24	.031				( +				
25	.041				<u> </u>				
26	.100				( ) •				
27	.123				XXX				
28	.121				XXX				
29	.104				III				
30	.111				III				
31	.127			+ I					
32	.082				11÷				
33	.021				 ▼ +				
34	011				÷				
	013			+ Ī					
26	021			+ 11					



NUMBER OF OPSERVATIONS	:	708
MBAN OF THE (DIFFERENCE) SERIES	:	
STANDARD ERROR OF THE MEAN	:	.0331
T-VALUE OF MEAN (AGAINST ZERC)	=	4.5841

1- 12 ST.E.												
13- 24	03	.01	.01	0.0	05	.04	.03	.01	02	.08	03	04
ST.E.	.04	.04	.04	.04	.04	. 04	.04	.04	.04	.04	.04	.04
25- 36	.03	.10	0.0	.01	0.0	.03	.03	04	05	03	# 1 1 - •	0.0
ST R	n A	.04	D.E	0.4	0.4	0.4	D.	0.4	D.A	2.1	D.E	0.4



## PLOT OF PARTIAL AUTOCORRELATIONS

LAG	-1.08 CORR. +	645	.0 +	.£	. 4	.5	. 8	1.0
			:					
1	.633		+ IX+X	. X X X X X	(XXXX)	LXXX		
5	.024		+ IX+					
3 4	.042		+ []+					
	1:1.		+ IX+					
5	.004 061		+ I +					
3			<b>XXI</b> +					
	018		+ I + + I +					
8 9	.015 056							
	029		+ X					
10 11			XXI +					
	044		+XI +					
12	051		+XI +					
13	031 .014		+ I #					
14 15	.014		† I †					
16	001		: 1 † + [ +					
10 17	001 053		† <u>1</u> † +∑ĭ +					
11	.036		*A. * + IX+					
16 19	.031		+ IX+					
20	.007		+ I +					
2:	024		+ XI +					
22	.079		+ []]					
23	030		; <u>18</u> 6					
24	035		+XI +					
25	.031		+ IX+					
26	.098		+ IXX					
27	.005		+ I +					
28	.013		. <u></u>					
29	.002		+ I +					
30	.023		+ IX+					
3:	.034		+ IX+					
32	039		+ XI +					
33	050		+ XI +					
3.4	028		+ X I +					
25	.965		+ I +					
38	-,002		. <u></u> .					

## VANDADIE IS SOLAN ADM

NUMBER OF CESTAVATIONS	:	<b>*</b> ;?
MEAN OF THE COOPPERENCES SERVES	2	438,000
STANDARD FREDE OF THE MEAN	:	4,045
THVALUE OF MEAN CAGAINST DESC'	2	

#### AUTOCOBREDATIONS

1- 12 ST.E.												
13- 24 ST.E.	.17		.09									
25- 36 ST.E.	.21	. 29	.30	.25	.17	.10	.05	00	06	-,19	1!	09

## FILT OF AUTOCORRELATIONS

LAG		.: 3 2. 0. 2 4 3 8
2A3		
1	.778	+ IX+XXXXXXXXXXXXXXXXX
2	.545	+ [XX+XXXXXXXXXXX
3	.365	+
2 3	.213	+ IXX+XX
ξ	.086	+ IXX+
€	.002	+ 1 +
7	053	+ ፮፻ •
8	064	+ % % ( ) +
9	021	+ 🏗 +
:0	. 353	• IX •
1:	.118	+ 1777
12	.164	+ IXX+X
. 0	.171	+ IXX+X
14	.125	+ IXXX
15	.093	• IXI+
3.1	.039	+ IX +
17	035	+ %1 +
18	097	+ X X X
19	:24	XXXI +
20	102	XXXI +
21	105	XXX: +
22	063	+XX: +
23	.001	+ I +
24	. 293	- IXX-
25	.213	+ IXX+XX
26	.288	+ IXXX+XXX
2:	.304	+ 1 X X X ÷ X X X X
28	.250	+ <b>IXXX+XX</b>
29	.174	+ IXXXX
30	.103	• ! X X X +
31	.051	+ IX +
32	018	+ I +
53	063	+ XXI +
34	035	• XX. •
3.5	110	+ 🛽 🕽 🕽 🔭 +
3.6	086	• XX -

#### FATE VARIABLE IS STUAR VINC

NUMERS OF OPSERVATIONS	:	*(:
MEAN OF THE (CIPPERSYJEC) SERIES	=	149,001
STANDARD ERROR OF THE MEAN	:	4.0045
T-VALUE OF MEAN (AGAINST ZEEL)	:	

1- 12 ST.E.												
13- 24	03	09	.05	06	06	01		.02	01	. 33	.05	.11
ST.E.	.04	.04	.04	.04	.04	.04	.04	.04	.04	.04	, 04	.04
25- 36	.16	.03	.01	09	02	.0:	.03	05	.0:	04	08	.03
ST.E.												

#### PLOT OF PARTIAL AUTOCORRELATIONS

LAG		8642 .0 .2 .4 .6	
		:	
4	.778	+ IX+XXXXXXXXXXXXXXXXXXXXXXXXXXXX	
2	150	XX+XI +	
3	017	+ 1 +	
4	051	+XI +	
5	088	XXI +	
6	.010	+ <u>I</u> +	
7	032	+XI_+	
8	.043	+ IX+	
9	300.	+ IXX	
10	.087	+ IXX	
11	.037	• IX-	
12	.029	+ IX+	
13	034	+XI +	
14	087	XXI +	
15	.048	+ 12+	
16	064	XXI +	
17	055	÷X _ +	
18	014	+ [ +	
19	.007	+ 1_+	
20	.031	+ 111+	
21	915	+ [ +	
22	.031	+ IX+	
23	.053	+ IX+	
24	.111	+ IX+X	
25	.159	+ IX+XX	
26	.030	+ IX+	
27	.007	+ I +	
28	093	XX +	
29	019	+ <u>I</u> +	
30	.012	+ I +	
31	.026	+ IX+	
32	049	+XI +	
33	.012	+ I +	
34	037	+XI +	
3.E	082	XXI -	
3.6	.015	+ <u>:</u> Y•	

# Appendix H: CCFs of the 70 Day Transfer Function Models

													Page
CCF	for	IMF $B_z$ and	SEESSD				•				•	•	197
CCF	for	$IMF B_z$ and	SEEI		•		•					•	200
CCF	for	IMF Bz and	SEEII			•	•						203
CCF	for	IMF Bz and	SEEIII		•	•	•	•				•	206
CCF	for	IMF Bz and	SEEIV	•	•					•		•	209
CCF	for	Solar Wind	and SEESSD		•							•	213
CCF	for	Solar Wind	and SEEI .		•	•			•			•	216
CCF	for	Solar Wind	and SEEII	•	•				•			•	219
CCF	for	Solar Wind	and SEEIII									•	222
CCF	for	Solar Wind	and SEEIV									•	225

## Appendix H: CCFs of the 70 Day Transfer Function Models

ARIMA VARIABLE IS ABOM.
ABOR = '(1)'.
CENTERED./

THE COMPONENT HAS BEEN ADDED TO THE MUDEL THE CURRENT MODEL HAS OUTPUT VARIABLE = ABZM INPUT VARIABLE = NOISE

ESTIMATION RESIDUALS = RX.
METHOD IS CLS./

ESTIMATION BY CONDITIONAL LEAST SQUARES METHOD

RELATIVE CHANGE IN BESIDUAL SUM OF SQUARES LESS THAN . 1000E-04

SUMMARY OF THE MODEL

OUTPUT VARIABLE -- ABZM INPUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFERENCES

ABEM RANDOM REMOVED 1- 70

PARAMETER VARIABLE TYPE FACTOR ORDER ESTIMATE ST. ERE.
T-RATIO

1 ABZM AR 1 1 .2982 .1141

1 ABZM AR 1 1 .2982 .114 2.61

RESIDUAL SUM OF SQUARES = 162.838822 DEGREES OF FREEDOM = 68 RESIDUAL MEAN SQUARE = 2.394689

FILTER VARIABLE IS SEBSSD.
RESIDUALS = RY./

RESIDUAL MEAN SQUARE = 39781.905604

VARIABLE SEESSO IS FILTERED, RESULTS ARE STORED IN VARIABLE BY

CCF VARIABLES ARE RX, RY. MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 69

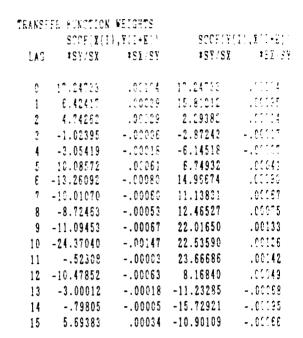
CORRELATION OF BY AND BY IS .12

CROSS CORRELATIONS OF RX (I) AND RY (I+E)

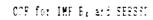
13-15 -.02 -.01 .04 ST.E. .13 .13 .14

CROSS CORRELATIONS OF BY (I) AND BX (I+K)

. 13- 15 -.09 -.12 -.08 ST.R. .13 .13 .14



WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD EREOR OF X(I), AND SY THE STANDARD ERROR OF Y(I)



## PLOT OF CROSS CORRELATIONS

	-1.0	864	:	2.0	. 2	. 4	, f	, ધ	1.0
LAG	COBE. +	-+	-+	- <del> </del>			+	-+	+
-15	085	+		XXI	÷				
-14	122	+		XXX:	•				
-13	087	+		XX!	+				
-12	.063		Ļ		+				
-11	.184		+	IXX	ZZZ+				
-10	.175		÷	III	¥				
- <u>9</u>	.171		÷		77 +				
-8	.097		+	IXX					
- ?	.085		÷	II					
-6	.116		ŧ	IX					
-5	.052		ŧ	IX	+				
-4	048		÷	ΣΪ	÷				
- 3	022		ŧ	X I					
-2	.018		÷	I	÷				
-1	.123		ŧ		i I +				
0	.134		ŧ		<b>Y</b> +				
	.050		÷	ΙÏ	+				
1 2 3	.037			IX	+				
3	008		ŧ	7	+				
4	024		÷	<b>X</b> :					
£	.078		÷	! Y Y					
€	103		÷	XXXI	÷				
7	078		ŧ	XXI	+				
8	068		+	XXI	÷				
9	086		+	XXI	+				
10	189		<b>+ X</b>	XXXX:	÷				
11	004		+	I	7				
12	081		+	XXI	+				
13	023	+		XI	+				
14	006			I	÷				
15	.044	+		II	+				



FILTER VARIABLE IS SEEL.
RESILIALS = RY./

BESIDUAL MEAN SQUARE = .998624

VARIABLE SEEL IS FILTERED, RESULTS ARE STORED IN VARIABLE BY

CCF VARIABLES ARE EX. BY. MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 69

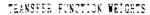
CORRELATION OF RY AND RY IS .21

CROSS CORRELATIONS OF RE (I) AND RY (I+E)

13- 15 -.13 -.09 .03 ST.E. .13 .13 .14

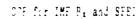
CROSS CORRELATIONS OF RY (I) AND RX (I+X)

10-15 -.14 -.19 -.13 ST.E. .13 .10 .14



	SCOF(X(I	),Y(I+E))	STOFT	(I),X(I+E')
LAC	ISY/SY	#SX/SY	18Y/8X	*SX/SY
C	.13706	.32951	.13700	.32950
1	.00813	.01955	.12438	.29905
2	03207	07711	00058	00140
3	03019	07259	01555	03738
4	.02199	.05289	00975	01345
5	.07196	.17303	.02308	.05548
6	.01973	.04743	.05624	.13524
7	01563	03758	.05928	.14254
8	.02169	.05217	.05720	.13778
9	05234	12587	.11599	.27890
10	12414	29851	.14433	.34820
1:	03319	07982	.13744	.33049
12	01282	03082	.02401	.05772
13	08438	20291	09298	22359
14	05672	13640	12448	29929
15	.02156	.05184	08815	20717

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD ERROR OF X(I), AND SY THE STANDARD ERROR OF Y(I)



## FIOT OF CROSS CEREELATIONS

		864							
LAG	CORE. +			<del>†</del> † -		· <b>+ -</b>	- •	-+	•
-15	134	+	XXX	!	+				
-14	193	+	XXXXX		+				
-13	144	· •			+				
-12	.037			ΪX					
-11	.213			IXXXXX	+				
-10	.225			IXXXX					
- 9	.180			IXXXX					
-8	.089			IXX	+				
-7	.092	,		IXX	÷				
-6	.087			IXX	+				
	.036			IX	ŧ				
- 4	015			I	†				
-3	024			Ī	· +				
-2	001			Ī	· •				
	.193			IXXXX					
Ô	.213			IXXXXX					
i	.013			I	ļ.				
2	050			Ī	+				
3	047			7	÷				
4	.034			IX	, +				
5	.112			IXXX	+				
6	.031			IX	†				
7	024			I	†				
8	.034			IX	_				
9	081		• II		+				
10	192		, , , , , , , , , , , , , , , , , , ,		Ŧ <del>ļ</del>				
11	051								
			+ <u>Y</u>		+				
12	020		+		<del>!</del>				
13	131	<del>†</del>			•				
14	088	<b>†</b>	XX		÷				
15	.030	<u> </u>		IX	*				

PILTER VARIABLE IS SERI!.

BESIDUALS = RY./

RESIDUAL MEAN SQUARE = .016699

VARIABLE SEET IS FILTERED, RESULTS ARE STORED IN VARIABLE BY

CCF VARIABLES ARE RX, RY. MAXLAG IS 15./

BFFECTIVE NUMBER OF CASES = 69

CORRELATION OF BY AND BY IS .13

CROSS CORRELATIONS OF RY (I) AND RY (I+K)

1- 12 .04 -.01 -.02 .07 .05 .02 -.04 .12 -.09 -.21 -.29 .30

ST.E. .12 .12 .12 .13 .13 .13 .13 .13 .13 .13 .13 .13

13- 15 -.17 -.08 -.14

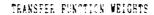
ST.B. .13 .13 .14

CROSS CORRECATIONS OF RY (I) AND RX (I+E)

1- 12 .04 -.04 -.09 .01 .12 .09 .05 .08 .12 .16 .16 -.02

13- 15 -.14 -.15 -.09

ST.E. .13 .13 .14



		[],Y(I+E]]		([],X[[+E])
LAG	*SY/SX	\$8X '8Y	*SY/SY	*SX/SY
C	.01597	2.24640	.01565	2.24640
1	.00295	.42242	.00013	.44441
5	00087	12533	00000	43693
3	00169	24158	00745	-1.06813
4	.00606	.88849	.00088	.12557
5	.00383	.54862	.00981	1.37729
6	.00207	.29638	.00769	1 10007
7	00308	44200	.0045~	.65564
8	.01035	1.48312	.00830	.90272
9	00769	-1.10281	.01008	1.44478
10	01783	-2.55647	.01363	1.95045
11	02406	-3.44902	.01302	1.85645
12	.02707	3.88153	00208	29774
13	01438	-2.06086	01170	-1.67688
14	00671	96248	01228	-1.75830
15	01135	-1.62682	0077€	-1.11309

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD ERROR OF X(I), AND SY THE STANDARD ERROR OF Y(I)

# COP for IMP E. and FIELD

# FLOT OF CROSS CORRELATIONS

	-1.08	64 -	. :	. C	-	. 4	. {	ç	
LAG	CORR. :	+	•	+		<b>.</b>	+		· - ÷
				I					
- : 5	093	+	XX	<u> </u>	+				
- 14	147	+	XXXX	Ī	•				
-13	140	+	XXX?	I	÷				
-12	025	+	Y	Ī	•				
-11	.156	+		XXXX	•				
-10	.163	+		IXXXX	÷				
- 9	.121	+		IXXX	1				
- 8	.075	+		XX	+				
-?	.055	+		IX	•				
-€	.092	+		XX	÷				
-5	.115	÷		IXXX	÷				
- 4	.010	÷		I	•				
- 3	089	+	XX	Ţ	+				
-2	036	+	X	Ţ	•				
-1 0 1 2 3	.037	<b>*</b>		Įχ	÷				
9	.188	•		IXXXI					
1	.035	+		ΙX	٠				
2	010	•			4				
3	020	+	X	•	•				
4	.073	•		ĮΫ́Υ	•				
5	.046	•		ĬΪ	•				
; ?	.015	+		Y	•				
	027	+	X	<u> </u>	•				
8	.124	+		XXX	•				
9	092	+	XX	-	•				
10	214	+1	<u> </u>	•	•				
11	288	X+2	XXXXX	÷	÷				
12	.324	+		IXXXX	<b>-2</b> 1				
13	172	+	XXXX	Ţ	•				
14	080	+	XX	I	÷				
15	136	+	XXX	7	÷				

FILTER VARIABLE IS SEBIII.

RESIDUALS = RY. 4

RESIDUAL MEAN SQUARE = .060225

VARIABLE SEELLI IS FILTERED, RESULTS ARE STORES IN VARIABLE BY

CCF VARIABLES ARE BX, RY. MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 69

CORRELATION OF BE AND E7 IS .17

CROSS CORRELATIONS OF BX (I) AND RY (I+E)

1- 12 .02 -.01 -.04 .06 .08 -.03 -.07 .05 -.10 -.12 -.18 .18

ST.E. .12 .12 .12 .13 .13 .13 .13 .13 .13 .13 .13

13- 15 -.12 -.11 -.04

ST.E. .13 .13 .14

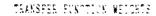
CROSS CORRELATIONS OF RY (I) AND BY (I+E:

1- 12 .19 -.05 -.14 -.03 .10 .13 .02 .08 .10 .22 .18 0.0

ST.E. .12 .12 .12 .13 .13 .13 .13 .13 .13 .13 .13 .13

13- 15 -.15 -.20 -.10

ST.E. .13 .13 .14



	8002181	[',V']+F	S0.15	Υ(1', <b>Σ</b> '1'+£''
1.4.0		*SY SY		7 15% SY
C	.00168	:1.35831	.66186	17.35507
:	.01116	1.70932	.00180	10,80896
2	00011	-1.05671	00052	-5.60809
3	00042	-4.42532	01134	-14.26331
4	.00059	6.33617	00028	-3.00017
5	.00055	5.87743	30000.	10.22470
6	00027	-2.88377	.00130	13.91736
7	00064	-6.86513	.00017	1.83434
8	.00048	5.06501	.00076	8.05138
g	00100	-10.68439	.00096	10.27352
10	00162	-17.24604	.00217	23.11200
11	00178	-18.97704	.00175	18.60925
12	.00179	19.07811	00003	-,30053
13	00121	-12.87591	00142	-15.09815
14	00107	-11.40583	00189	-20.16544
15	00042	-4.50083	00100	-10.64196

WHERE x(i) is the first series, y(i) the second series, sx the standard error of x(i), and sy the standard error of y(i)

KSSSSSS TOSSSSS TO A PROPERTY OF THE PROPERTY

S.W.

# COF for IMP By and SERIII

# FLOT OF CROSS CORRELATIONS

	-1.0	8 -,6 -,4 -	.2 .0	.2	. 4	Ţ.	. [	
LAG	COEE. ++		+		-+	• •	-+	• •
			:					
-15	103	+		•				
-14	195	+ :	IXXXXI	+				
-13	146	+		•				
-:0	003	+	I	+				
	.180	+	IXX	X				
-10	. 224	+	ΙXΧ	XXXX				
- 9	.100	+	IXX	+				
-8	.078	+	IXX	•				
- ?	.018	+	I	÷				
-£	.134	+	IXX	<u>y</u> •				
- 5	.099	+	ΙXX	•				
- 4	029	ŧ	<b>X</b> I	+				
- 3	138	+	XXXI	+				
-2	054	+	XI.	+				
-1	.192	+	IXX	X				
0	.168	+	IXX	77 +				
1	.017	÷	I	+				
2	010	+	I	+				
3	043		XI	+				
4	.061	+	: X X	+				
5	.057	+	IΣ	+				
6	028	+	XI	÷				
?	330	+	XXI	+				
8	.049	+	IX	÷				
9	103	+	XXXI	÷				
10	167	÷	IXXXI	+				
11	184	+	IXXXXI	+				
12	.185	+	IXX	<b>777</b> +				
13	125	+	XXXI	+				
14	110	+	XXXI	+				
15	044	+	ΧI	+				

FILTER VARIABLE IS SEELV.
RESILUALS = BY.

RESIDUAL MEAN SQUARE = .0000000

VASIABLE SEELV IS FILTERED, BESULTS ARE STIREL IN VARIABLE BY

CCF VARIABLES ARE RX, RY. MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 69

CORRELATION OF EX AND RY IS .20

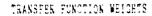
CROSS CORRELATIONS OF RX (I) AND RY (I+E)

13- 15 -.12 -.09 -.04 ST.E. .13 .13 .14

CROSS CORRELATIONS OF BY (I) AND RY (I-E)

13- 15 -.16 -.19 -.10 ST.E. .13 .13 .14

63/656566 Proceeds Colleges Spreader Machineses Medicales Services Spreader



	SCCF(X)	I],Y(I+E))	SCOF	Y(I),X(I+E')
LAG	*SY/SX	ISX/SY	\$\$Y/\$	X ISXISY
e	.00219	19.01547	.00219	19.01547
1	.00005	.39192	.00249	21.66936
2	.00004	.33089	00057	-4.96128
3	00073	-6.35314	00169	-14.69445
4	.00085	7.40493	00044	-3.82058
5	.00094	8.13439	.00107	9.30792
6	00029	-2.50574	.00115	10.03192
7	00098	-8.52418	.00007	.62559
8	.00029	2.55164	00008	65329
9	00124	-10.73600	.00098	8.50349
10	00129	-11.21067	.00209	18.15729
11	00154	-13.38780	.00182	15.81787
12	.00149	12.91103	00011	91728
13	00126	-10.96165	00109	-14.73027
14	0009?	-8.42227	00203	-17.64997
15	00039	-3.39267	00105	-9.12247

WHERE x(i) is the first series, y(i) the second series, sx the standard ereor of x(i), and sy the standard error of y(i)

## COF for IMP By and SEETY

## PLOT OF CROSS CORRELATIONS

	-1.0	854	-	. 2	) .	n L	. ;	. £	, :	
LAG	COES. +	• • •	- <b>-</b> -				÷	· • • • -	- <b>•</b> -	- +
,,	0.00			1						
-15	098		+			÷				
-14	189			ZZZZZ		+				
-13	158		÷	XXXXI.		•				
-12	010		•	Ī		•				
-11	.170		+		XXXX					
-10	.195		ŧ		ZZZZ					
- 9	.091		ŧ			ł				
-8	007		ŧ	I		ŧ				
- 7	.007		÷			+				
-8	.108		÷			ŧ				
- 5	.100		÷	I	XX	ŧ				
- 4	041		÷	X I		t				
- 3	158		+	IXXXI		÷				
- 2	053		÷	ΙI		ļ				
-1	.232		+	I	XXXXX	X				
0	.204		÷	I	IXXX	÷				
1	.004		+	I		ļ				
2	.004		÷	I		ţ				
3	068		÷	M		Ļ				
4	.079		ŧ		X Y	Ļ				,
5	.087		÷	I		ŧ				
6	027		+	XI		ŧ				
7	091		÷	ZZI		•				
8	.027		+			•				
9	115		ŧ			· •				
10	120		+	YYYI						
11	144		: +	XXXXI		•				
12	.138		τ †			- <del> </del>				
12	118		Ť		4.4					
14			† †			+				
	090			XXI		+				
15	038		ŧ	ΪΪ		÷				

ARIMA VARIABLE IS AV.

AROR = '(1''.

CENTERED./

THE COMPONENT HAS BEEN ADDED TO THE MODEL

THE CURRENT MODEL HAS
OUTPUT VARIABLE = AV
INPUT VARIABLE = NGISE

BSTIMATION RESIDUALS = EΣ.

METHOD IS CLS./

BSTIMATION BY CONDITIONAL LEAST SQUARES METHOD

RELATIVE CHANGE IN EACH ESTIMATE LESS THAN .1000E-03

SUMMARY OF THE MODEL

OUTPUT VARIABLE -- AV
INPUT VARIABLES -- NOISE

VARIABLE VAR. TYPE MEAN TIME DIFFERENCES

AV RANDOM REMOVED 1- 71

PARAMETER VARIABLE TYPE FACTOR ORDER ESTIMATE ST. ERR.

T-RATIO

1 AV AR 1 1 .8327 .6662

12.57

RESIDUAL SUM OF SQUARES = 267796.403915 DEGREES OF FREEDOM = 69 RESIDUAL MEAN SQUARE = 3881.107303

FILTER VARIABLE IS SEESSI.

RESIDUALS = RY./

RESIDUAL MEAN SQUARE = 20230.824039

VARIABLE SEESSY IS BILTERED, RESULTS ARE STORED IN VARIABLE BY

# CCF VARIABLES ARE RX, RY. MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 70

-.06 .08 -.08

.13 .13 .13

13- 15

ST.E.

CORRELATIO	O K	GF	RX		AN	D RY	IS	.::	
CROSS COR	RELATION	IS OF	£X	{	I; AN	D RY	(I+E)		
							0315 .13 .13		
13- 15 ST.E.									
CROSS COR	RELATION	IS OF	RY	(	I) AK	D RX	(I+E)		
							.0104 .13 .13		

#### TRANSFER FUNCTION WEIGHTS

	SCCF(X(I	),Y(I+E))	SCCF (Y	(I), Y(I+E
LAG	*SY/SX	*SX/SY	≠SY/SX	\$\$X/\$Y
0	. <u>17773</u>	.04111	.27773	.04111
1	.60608	.08971	14008	02073
£	.66813	.03889	.02108	1:000.
3	.75781	.11218	.09824	.01424
Ą	08774	01299	.12262	.01815
5	.41413	.06130	.06828	.01011
€	.10190	.01508	26208	03879
7	47299	07001	.04831	.00715
8	22486	03328	.01420	.00210
9	40081	05933	11429	01692
10	34056	05041	28235	04179
11	10616	01571	09514	01408
12	07602	01125	14287	02115
13	18833	02787	15903	02354
14	09610	01422	.20099	.02975
15	03737	00553	20725	03968

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECONT SERIES, SX THE STANDARD BERGE OF X(I), AND SY THE STANDARD BERGE OF Y(I)



## COP for Solar wind and SEESSI

## PLOT OF CROSS CORRELATIONS

		-		. 2	. 4	.€	.;	1.0
LAG	CORR. ++	-+	· <del>i</del> I	+	• • • •	- +	+	+
-15	080	+	XXI	÷				
-13	.077	† †	IXX	† †				
-13	061	<b>*</b>	XXI					
	055	+	XI	•				
	037	, †	XI	÷				
	109	· ·	XXXI	, T				
-10 -9	044	+	XI.					
-8	.005	•	I					
-0 -7	.019	· · · · · · · · · · · · · · · · · · ·	Ï	+				
- f	101	+	YYYI	T .				
-6 -5	.02€	† +	IX	+				
- 0 - 4	.047	†	IX	+				
- 4 - 3	.037	+	IX	1				
- 3 - 2		†	Ţ	τ +				
-2 -1	.008	+	XI	+ +				
	054 .107	7	IXX					
0	.233	+		XXXX				
! 2				XXXX				
	.257	+		X				
3	. 292							
4	034	+	XI	† • • •				
5	.159	+						
6	.039	+	II	*				
?	182		IXXXXI	+				
8	08?	+	XXI	+				
9	154	+	MAMI	+				
10	131	+		+				
11	041	+		+				
12	029	+		÷				
13	072	+		+				
14	037	+	XI	÷				
15	014	+	Ĭ	•				





FILTER VARIABLE IS SEEL.
RESIDUALS = EV./

RESIDUAL MEAN SQUARE : 1.031222

VARIABLE SEEL IS FILTERED, BESULTS ARE STORED IN VARIABLE BY

CCF VARIABLES ARE RX, RY.
MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 70

CORRELATION OF BY AND BY IS -.20

CROSS CORRELATIONS OF RX (I) AND RY (I+K)

13- 15 .03 -.09 -.02 ST.E. .13 .13 .13

CROSS CORRELATIONS OF BY (I) AND BY (I+E)

1- 12 -.09 -.05 .07 .10 -.17 .10 .06 -.06 -.01 .01 -.01 ST.E. .12 .12 .12 .12 .12 .13 .16 .16 .16 .16 .16 .16 .16 .16

13- 15 -.06 .04 -.02 ST.E. .13 .13 .13



#### TRANSFER FUNCTION WEIGHTS

LAC	SCCF(X( *SY/SX	I), Y(I+K); *SX/SY		::,,X(I+E'); *SX/SY
0	00419	-9.98864	01419	-9.98884
1	.00418	9.95424	00194	-4.61691
2	.00646	15.37227	00105	-2.50526
3	.00436	10.37493	.00152	3.62023
4	.00206	4.89085	.00215	5.11054
5	00520	-12.37819	00345	-8.21165
€	.00439	10.44747	.00198	4.70534
?	.00343	8.15340	.00127	3.03118
8	00436	-10.36712	00091	-1.44978
9	00276	-6.56623	00118	-2.80621
10	00176	-4.18948	00234	-5.55918
11	00097	-2.31229	.00528	.66033
12	.00128	3.04568	00017	40553
13	.00064	1.51892	00120	-2.86510
14	00183	-4.49396	.00090	2.15073
15	00047	-1.11263	00048	-1.14934

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD BRROR OF X(I), AND SY THE STANDARD BEROR OF Y(I)



#### FLOT OF GROSS CORRELATIONS

	-1.08			·£ .	3. }	.5 :
LAG	CORR. ++	++				
-:5	024	÷	I XI			
-14	.044	· •	IX	+		
	059	•	XI.			
	008	+	Ţ	† †		
-11	.014	+	Ī	· ·		
-10	114	•		· •		
	058	•	XI	+		
-8	030		XI	+		
- ?	.062	+	IXI			
-6	.098	· +	IXX	•		
- 5	168		XXXXI	· +		
- <b>4</b>	.105	+	IXXX			
- 3	.074		IXX	+		
-2	051	· •	XI	· •		
-1	095		XXI	· •		
Ô	204		IXXXXI	· +		
1	.204	+	IXXX			
2	.315			XX+XX		
	.213	+	IXXX			
3 4	.100		IXXX			
5	254	<b>7</b> ,	XXXXI			
6	.214		IZZX			
?	.167		IXXX			
8	213		XXXXI			
9	135	+		4		
10	086	•	XXI			
11	047	<b>.</b>	XI.			
12	.062	· •	IXX			
13	.031		IX	÷		
14	032	•	XXI			
15	023	÷	XI	4		
10	. 5 2 0	•	V.1			



FILTER VARIABLE IS SEEIL.
RESIDUALS = RV.;

RESIDUAL MEAN SQUARF = .024945

VARIABLE SBELL IS FILTERED, RESULTS ARE STOREL IN VARIABLE BY

COF VARIABLES ARE RX, RY.
MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 70

CORRELATION OF BY AND BY IS -.17

CROSS CORRELATIONS OF RX (I) AND BY (I+E)

1- 12 -.06 .33 .26 .05 .03 -.26 .28 .14 -.53 -.05 -.02 -.00 ST.E. .12 .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13

12- 15 .09 0.0 -.11 ST.B. .13 .12 .13

GROSS CORRELATIONS OF BY (I) AND RX (I+E)

1- 12 -.06 -.03 .14 -.13 .01 .10 -.02 -.01 -.03 -.05 -.04 .02 ST.E. .12 .12 .12 .12 .12 .13 .13 .13 .15 .13 .13 .15

13- 15 -.06 .01 -.01 ST.E. .13 .13 .13

## TRANSFER PUNCTION WELDETS

	SCCF[X[],F(]+E[]	3275(11111), X 1+1111
LAC	187/87 187/87	13Y/SE 157/3Y
ŗ	00044 -09.93091	0004409.90.91
•	00014 -20.14014	00014 -22.21000
ď	.00090 103,85795	-,00009 -10.67845
3	.00008 100.83408	.00036 56.66090
4	.00012 18.29070	00000 - <b>5</b> 2.03500
Ę	.00008 12.57259	0 000 -01.000. .00002
E	00067 -100.41849	
•		.00026 40.05864
7	.00072 111.57536	00005 -8.10037
8	.00034 50.59599	00003 -4.34291
9	00085 -131.85773	00009 -13.36425
10	00013 -19.47069	00012 -18.26139
11	0000E -9.14531	00009 -14.51965
12	00007 -10.65153	.00005 7.62975
12	.00023 35.65978	00015 -22.81577
14	.00001 1.41485	.00004 5.59812
15	00027 -42.23735	00004 -5.69105

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECOND SERIES, SX THE STANDARD ERROR OF X(I), AND SY THE STANDARD ERROR OF Y(I)

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FILTER VARIABLE IS SEEIII.

BESIDUALS = RY./

RESIDUAL MEAN SQUARE = .000000

VARIABLE SEBILI IS FILTERED, RESULTS ARE STORED IN VARIABLE BY

CCF VARIABLES ARB RY, RY.
MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 70

CORRELATION OF BY AND BY IS -.15

CROSS CORRELATIONS OF RX (I) AND RY (I+E)

1- 12 .09 .48 .29 .11 -.01 -.19 .24 -.01 -.35 -.08 -.04 -.04

ST.E. .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13

13- 15 .04 -.04 -.08

ST.E. .13 .13 .13

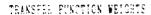
CROSS CORRELATIONS OF RY (I) AND BX (I+E)

1- 12 -.07 -.0: .11 -.06 -.06 .01 -.04 -.02 -.06 -.08 6.0

ST.E. .12 .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13

13-15 -.05 .01 -.02

ST.E. .13 .13 .13



	SCCF(X(I),Y(I+E))	SCCF(Y(I ,X(I-E))
040	*SY/SX *SX,SY	*SY/SX *SX,SY
(i	00004 -557.22820	08864 -557.22823
1	.00002 345.03445	00002 -258.34425
2	.00012 1752.94110	00000 -37.77847
3	.00008 1087.46988	.00003 416.53924
4	.00003 417.24312	00001 -210.61234
5	00000 -20.24466	00002 -235.22776
E	00005 -716.83016	.00001 213.51534
?	.00006 912.54635	.00000 42.79133
8	00000 -32.34764	00001 -162.04084
9	00009-1321.47477	00001 -73.60908
10	00002 -298.71936	00002 -228.34888
11	00001 -139.99961	00002 -301.17763
12	00001 -135.18685	00000 -15.76826
13	.00001 151.29790	00001 -191.65939
14	00001 -162.07944	.00000 39.86271
15	00002 -284.91273	00000 -68.05244

WHERE X(I) IS THE FIRST SERIES, Y(I) THE SECONT SERIES, SX THE STANDARD ERROR OF X(I), AND SY THE STANDARD ERROR OF Y(I)

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#### COF for Solar Wini and SEEIII

#### PLOT OF CROSS CORRELATIONS

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8	009			ŧ	Ţ		+				
9	348			X	ZXXXI		+				
10	079			+	XXI		÷				
11	037			+	<b>X</b> I		ŧ				
12	036			+	ΣI		÷				
13	.040			+	I	ĭ	÷				
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FILTER VARIABLE IS SEELV.
RESIDUALS = BY./

RESIDUAL MEAN SQUARE = .000083

WARRABLE SERIV IS FILTEPED, RESPUTS ARE STORED IN WARRABLE BY

CCF VARIABLES ARE RX, RY.
MAXLAG IS 15./

EFFECTIVE NUMBER OF CASES = 70

CORRELATION OF BX AND RY IS -.12

CROSS CORRELATIONS OF RX (I) AND RY (I+E)

ST.E. .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13

13-15 .01 -.04 -.06 ST.E. .13 .13 .13

CROSS CORRELATIONS OF RY (1) ASD RX (1+E)

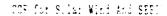
1- 12 -.09 -.08 .18 -.04 -.07 .03 -.01 -.01 -.01 -.09 -.08 -.02 ST.B. .12 .12 .12 .12 .12 .13 .13 .13 .13 .13 .13 .13

13- 15 0.0 .03 -.04 ST.E. .13 .13 .13

#### TRANSFER FUNCTION WEIGHTS

	SCCF(X(I),Y(I+E))	SCCF:Y(I),X(I+E);
LAG	*SY/SX *SX/SY	*SY/SX *SX/SY
e	00003 -437.02456	00093 -427.02456
1	.00004 551.92268	00002 -334.21850
2	.00011 1537.17671	00000 -206.28850
3	.00007 967.08784	.00004 585.73900
4	.00003 347.53209	00001 -140.69058
£	00000 -27.79294	00002 -262.10206
٤	00001 -187.72069	.00001 124.30921
?	.00004 542.41028	00000 -22.77830
8	00092 -319.59358	00000 -19.33228
9	00008-1047.55819	00001 -120.98443
10	00002 -316.49879	00002 -317.51343
11	00001 -175.76318	00052 -235.75233
12	00000 -5.72133	00001 -84.05070
13	.00000 38.24121	00000 -2.22640
14	00001 -151.77093	.00001 110.36804
15	00002 -207.36667	00001 -157.72481

WHERE X(1) IS THE FIRST SERIES, Y(1) THE SECOND SERIES, SX THE STANDARD ERROR OF X(1), AND SY THE STANDARD BREOR OF Y(1)



#### PIOT OF CROSS CORRELATIONS

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-11	064	+	XXI	÷		
-10	087	+	XXI	+		
- 9	033	+	XI	+		
- 8	005	+	I	+		
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  Engineering, Air Force Institute of Technology (AU),
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#### VITA

Major Michael P. Halpin was born on 18 November 1951 in Albuquerque, New Mexico. He graduated from high school there in 1969 and briefly attended the University of New Mexico before accepting an appointment to the United States Air Force Academy in June 1970. In June 1974, he graduated from USAFA with a Bachelor of Science degree in Civil Engineering and received a commission in the USAF. Subsequent to this, Major Halpin attended pilot training at Reese AFB, Texas where he was awarded his wings in September 1975. Major Halpin flew heavy aircraft including the KC-135 and E-3A AWACS. While stationed at Tinker AFB, Oklahoma, he completed the requirements for another Bachelor of Science degree, this time in Petroleum Engineering in May 1983. Major Halpin entered the School of Engineering, Air Force Institute of Technology, in May 1985.

Permanent address: 2423 9th Ave. Canyon, Texas 79015

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James Lange, Major, USAF	22b. TELEPHONE (Include Area Code) 513-255-6144	AFWAL/AARI
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SECURITY CLASSIFICATION OF THIS PA

This project used a Box and Jenkins time series analysis of energetic electron fluxes measured at geosynchronous orbit in an effort to derive prediction models for the flux in each of five energy channels. In addition, the technique of transfer function modeling described by Box and Jenkins was used in an attempt to derive input-output relationships between the flux channels (viewed as the output) and the solar wind speed or interplanetary magnetic field (IMF) north-south component,  $B_z^{\prime\prime}$ , (viewed as the input). The transfer function modeling was done in order to investigate the theoretical dynamic relationship which is believed to exist between the solar wind, the IMF B2, and the energetic electron flux in the magnetosphere. The models derived from the transfer function techniques employed were also intended to be used in the prediction of flux values.

The results from this study indicate that the energetic electron flux changes in the various channels are dependent on more than simply the solar wind speed or the IMF B2. Also, most of the time series models developed here (for both the individual energetic electron channels by themselves and those developed through transfer functions) were not suitable for use in prediction, since the standard error of the forecasts made using these models was unacceptably high. However, a few of the models did merit possible consideration for use in prediction of fluxes. These were the individual time series models for the 6.6 9.7 MeV channel. In addition, the transfer function models developed using the solar wind as an input and the 6.6 - 9.7 MeV channel as an output may be of possible use. channel containing electrons with energies between 9.7 - 16 MeV was also related to the solar wind via a transfer function with a reasonable forecast standard error. Finally, most of the transfer function models derived with the solar wind considered as the input to a given channel resulted in delay parameters of about 2 days between the input change in solar wind velocity and the observed output change in electron flux which supports findings from prior studies.

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